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OPEN Experimental solubility and thermodynamic modeling of empagliflozin in supercritical carbon dioxide

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The solubility of empagliflozin in supercritical carbon dioxide was measured at temperatures (308 to 338 K) and pressures (12 to 27 MPa), for the first time. The measured solubility in terms of mole faction ranged from 5.14 × 10⁻⁶ to 25.9 × 10⁻⁶. The cross over region was observed at 16.5 MPa. A new solubility model was derived to correlate the solubility data using solid-liquid equilibrium criteria combined with Wilson activity coefficient model at infinite dilution for the activity coefficient. The proposed model correlated the data with average absolute relative deviation (AARD) and Akaike's information criterion (AIC,), 7.22% and - 637.24, respectively. Further, the measured data was also correlated with 11 existing (three, five and six parameters empirical and semi-empirical) models and also with Redlich-Kwong equation of state (RKEoS) along with Kwak-Mansoori mixing rules (KMmr) model. Among density-based models, Bian et al., model was the best and corresponding AARD% was calculated 5.1. The RKEoS + KMmr was observed to correlate the data with 8.07% (correspond AIC, is - 635.79). Finally, total, sublimation and solvation enthalpies of empagliflozin were calculated.

List of symbols

A-D	New model constants
$A_1 - C_1$	Alwi-Garlapati model constants
$A_2 - C_2$	Bartle model constants
$A_3 - E_3$	Bian model constants
A_4 and B_4	Chrastil model constants
$A_{5} - E_{5}$	Garlapati-Madras model constants
$A_6 - C_6$	Kumar-Johnstone model constants
$A_7 - C_7$	Mahesh-Garlapati model constants
$A_8 - C_8$	Mendez-Teja model constants
$A_{9} - F_{9}$	Sodefian model constants
A_{10} and B_{10}	Reformulated Chrastil model constants
$A_{11} - F_{11}$	Tippana–Garlapati model constants
AARD	Absolute average relative deviation
Adj.R ²	Adjusted R ²
AIC	Akaike's information criterion
a _{ij}	EoS energy parameter
b _{ii}	EoS volume correction
Ć	Solubility in Chrastil model
Cp	Heat capacity
EoS	Equation of state
H _{sol}	Solvation enthalpy

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H _{sub}	Sublimation enthalpy
H _{Total}	Total enthalpy
H_2^m	Melting enthalpy of the solute
Mscf	Molecular weight of supercritical fluid
N	Number of data points
Р	Total pressure
P ^{sub}	Sublimation pressure
RK	Redlich-Kwong
P _r	Reduced pressure
P	Critical pressure
Q	Number of parameters of a model
R	Universal gas constant
R ²	Square of correlation coefficient
RMSE	Root mean square deviation
SSE	Error sum of squares
Т	Temperature
T _c	Critical temperature
T _m	Melting temperature
Tr	Reduced temperature
V ₂	Solubility in molefraction

Greek symbols

Δ	Difference
\hat{arphi}_i^S	Fugacity coefficient of the pure substance at saturation
$\hat{\varphi}_i^{ScCO_2}$	Solute fugacity in supercritical carbon dioxide (ScCO ₂)
$\overline{\omega}$	Acentric factor
ρ	Density
ρ_r	Reduced density
κ _{ii}	EoS mixing rule parameter
l_{ij}	EoS mixing rule parameter
$\dot{\lambda_{12}}, \lambda_{21}$	Wilson model parameters
γ_2^∞	Infinite dilution activity coefficient

Sub and superscripts

	-	-
exp		Experimental
cal		Calculated
1		Solvent (CO ₂)
2		Solute (drug)
с		Critical
m		Melting
r		Reduced

Supercritical carbon dioxide (ScCO₂) is a fluid above its critical point. It has physical properties (density, diffusivity, viscosity and surface tension) intermediate to that of gas and liquid^{1,2}. ScCO₂ has been used as a solvent in various process applications, because it has gas like diffusivity and liquid like density with low viscosity and surface tension^{1,3-5}. The major applications are in drug particle micronization, food processing, textile dyeing, ceramic coating, extraction and many more^{4,6-12}. Although, several supercritical fluids are utilized as solvent in process industry, ScCO₂ is the most desirable solvent^{8,13-17}. In general, phase equilibrium information is necessary to implement supercritical fluid technology (SFT)^{6,7,9}. The solubility is the basic information for the design and development of SFT. In literature, solubility of many drug solids in ScCO₂ is readily available¹⁸⁻³⁰, however, the solubility of *empagliflozin* has not been reported, therefore in this work for the first time, its solubility in ScCO₂ has been measured. This data may be used in the particle micronization process using ScCO₂. The molecular formula of *empagliflozin* is C₂₃H₂₇ ClO₇ and its molecular weight is 450.91. The chemical structure is shown in Fig. 1.

Empagliflozin is an inhibitor of sodium-glucose co-transporter-2 (SGLT2), the transporters primarily responsible for the re-absorption of glucose in the kidney. Further, it is useful in reducing the risk of cardiovascular death in adults with type 2 diabetes mellitus and cardiovascular disease³¹. Sufficient drug dosage is very essential for those treatments and this is achieved through a proper particle size. Therefore, the present study is quite useful in particle micronization using ScCO₂. Solubility measurement at each desired condition is very cumbersome and hence, there is a great need to develop a model that correlates/predicts the solubility³². Recent developments such as machine learning methods may be considered with the improvement of artificial intelligence prediction methods for the data correlation³³⁻³⁵. However, in general, the solubility models are classified into five types; however, only three are user friendly, and they are equation of state, density-based and mathematical models³⁶. Directly or indirectly all of them are derived based on thermodynamic frame work. The derived models make use of the basic concepts related to phase equilibrium criteria (solid–gas or solid–liquid), solvent–solute association theory, dilute solution theory, solution theory and Wilson model or any other model³⁷. In fact, most of the literature models correlate the solubility of the solid solutes in ScCO₂ quite well. A solid–gas equilibrium models need the



Figure 1. Empagliflozin chemical structure.

Compound	Formula	M _W (g/mol)	T _m (K)	λ_{max} (nm)	CAS number	Minimum purity by supplier
Empagliflozin	C23H27ClO7	450.9	426.1	276	864070-44-0	99%
Carbon dioxide	CO ₂	44.01			124-38-9	99.99%
DMSO	C ₂ H ₆ OS	78.13			67-68-5	99%

Table 1. Some physicochemical properties of the used materials.

critical properties and vapour pressure of the solute, while these properties are rarely available in literature, due to this, the group contribution methods are commonly used³⁸. On the other hand, the solid–liquid equilibrium (SLE) criterion requires an appropriate model for activity coefficient calculation. A recent study reveals that SLE model in combination with Van Laar activity coefficient model can be a simple approach in model development, but this method resulted in an implicit expression in terms of mole fraction^{38,39}. Therefore, there is a need to develop an explicit solubility model and hence, this task is taken up in this work.

The main motives of this study were in two levels. In the first level, *empagliflozin* solubility in $ScCO_2$ was determined and in the second level, a new explicit solubility model was developed based on solid–liquid equilibrium criterion in combination with Wilson activity coefficient model for the activity coefficient calculation.

Experimental

Materials. Gaseous CO_2 (purity > 99.9%) was obtained from Fadak company, Kashan (Iran), *empagliflozin* (CAS Number: 864070-44-0, purity > 99%) was purchased from Amin Pharma company, and dimethyl sulfoxide (DMSO, CAS No. 67-68-5, purity > 99%) was provided from Sigma Aldrich company. Table 1 indicates all the information about the chemicals utilized in this work.

Experiment details. The detailed discussion of the solubility apparatus and equilibrium cell has been presented in our earlier studies (Fig. 2)^{19,25,40,41}. However, a brief description about the apparatus is presented in this section. This method may be classified as an isobaric-isothermal method⁴². Each measurement was carried out with high precision and temperatures and pressures were controlled within ± 0.1 K and ± 0.1 MPa, respectively. For all measurement, 1 g of *empagliflozin* drug was used. As mentioned in our previous works, the equilibrium was observed within 60 min. After equilibrium, 600 µL saturated ScCO₂ sample was collected via 2-status 6-way port valve in a DMSO preloaded vial. After discharging 600 µL saturated ScCO₂, the port valve was washed with 1 ml DMSO. Thus, the total saturation solution was 5 ml. Each measurement was repeated thrice and their average values were reported. Mole fraction is obtained as follows:

$$y_2 = \frac{n_{drug}}{n_{drug} + n_{CO_2}} \tag{1}$$

where n_{solute} is the moles number of the drug, and n_{CO_2} is the moles number of CO_2 in the sampling loop. Further, the above quantities are given as:

n

$$_{\text{solute}} = \frac{C_s \cdot V_s}{M_s} \tag{2}$$

$$a_{\rm CO_2} = \frac{V_1 \cdot \rho}{M_{\rm CO_2}} \tag{3}$$

where C_s is the drug concentration in saturated sample vial in g/L. The volume of the sampling loop and vial collection are $V_1(L) = 600 \times 10^{-6} \text{ m}^3$ and $V_s(L) = 5 \times 10^{-3} \text{ m}^3$, respectively. The M_s and M_{CO_2} are the molecular weight of drug and CO₂, respectively. Solubility is also described as

$$S = \frac{C_S V_s}{V_1} \tag{4}$$

The relation between S and y_2 is



Figure 2. Experimental setup for solubility measurement, E1—CO₂ cylinder; E-2—Filter; E-3—Refrigerator unit; E-4—Air compressor; E-5—High pressure pump; E-6—Equilibrium cell; E-7—Magnetic stirrer; E-8—Needle valve; E-9—Back-pressure valve; E-10—Six-port, two position valve; E-11—Oven; E-12—Syringe; E13—Collection vial; E-14—Control panel.

$$S = \frac{\rho M_s}{M_{\rm CO_2}} \frac{y_2}{1 - y_2}$$
(5)

A UV–Visible spectrophotometer (Model UNICO-4802) and DMSO solvent were used for the measurement of *empagliflozin* solubility. The samples were analyzed at 276 nm.

Existing and new models and their correlations

In this section, the details of various solubility models are presented along with a new explicit solubility model.

Existing models. Alwi-Garlapati model (three parameters model)⁴³. It is one of the latest models for the solubility correlation. It is mathematically explained as

$$y_{2} = \frac{1}{\rho_{r}T_{r}} \exp\left(A_{1} + \frac{B_{1}}{T_{r}} + C_{1}\rho_{r}\right)$$
(6)

where $A_1 - C_1$ are model constants.

Bartle et al., model (three parameters model)⁴⁴. It is an empirical model and mathematically stated as:

$$\ln\left(\frac{y_2 P}{P_{ref}}\right) = A_2 + \frac{B_2}{T} + C_2(\rho - \rho_{ref}) \tag{7}$$

where $A_2 - C_2$ are model constants. From parameter B_2 , one can estimate sublimation enthalpy using the relation, $\Delta_{sub}H = -B_2R$, in which *R* is universal gas constant.

Bian et al., model (five parameters model)⁴⁵. It is an empirical model and mathematically formulated as:

$$y_2 = \rho_1^{(A_3 + B_3 \rho_1)} \exp\left(C_3 / T + D_3 \rho_1 / T + E_3\right)$$
(8)

where $A_3 - E_3$ are model constants.

*Chrastil model (three parameters model)*⁴⁶. It is a semi-empirical model and mathematically stated as:

$$c_2 = \rho_1^{\kappa} \exp\left(A_4 + \frac{B_4}{T}\right) \tag{9}$$

where κ , A_4 and B_4 are model constants.

In terms of mole fraction, it is written as⁴⁷:

$$y_{2} = \frac{(\rho_{1})^{\kappa-1} \exp\left(A_{4} + \frac{B_{4}}{T}\right)}{\left[1 + (\rho_{1})^{\kappa-1} \exp\left(A_{4} + \frac{B_{4}}{T}\right)\right]}$$
(9a)

*Garlapati–Madras model (five parameters model)*⁴⁸. It is a mathematical model and mathematically formulated as

$$\ln(y_2) = A_5 + (B_5 + C_5\rho_1)\ln(\rho_1) + \frac{D_5}{T} + E_5\ln(\rho_1 T)$$
(10)

where $A_5 - E_5$ are model constants.

*Kumar–Johnstone model (three parameters model)*⁴⁹. It is a semi empirical model and mathematically described as:

$$\ln(y_2) = A_6 + B_6\rho + C_6/T \tag{11}$$

where $A_6 - C_6$ are model constants.

*Mahesh–Garlapati model (three parameters model)*³⁹. It is one of the latest models. It is based on degree of freedom and mathematically stated as:

$$\ln(y_2) = A_7 + B_7 \rho_r T_r + C_7 \rho_r T_r^3$$
(12)

where $A_7 - C_7$ are model constants.

Mendez-Teja model (three parameters model)⁵⁰. It is a semi-empirical model and mathematically explained as:

$$T\ln(y_2 P) = A_8 + B_8 \rho + C_8 T \tag{13}$$

where $A_8 - C_8$ are model constants.

Sodefian et al., model (six parameters model)⁴⁰. It is a mathematical model and stated as:

$$\ln(y_2) = A_9 + \frac{B_9 P^2}{T} + C_9 \ln(\rho_1 T) + D_9(\rho_1 \ln(\rho_1)) + E_9 P \ln(T) + F_9 \frac{\ln(\rho_1)}{T}$$
(14)

where $A_9 - F_9$ are model constants.

Reformulated Chrastil model (three parameters model)^{47,51}. It is a semi-empirical model and mathematically explained as:

$$y_2 = \left(\frac{RT\rho_1}{M_{CF}f}\right)^{\kappa'-1} \exp\left(A_{11} + \frac{B_{11}}{T}\right)$$
(15)

where κ' , A_{10} and B_{10} are model constants.

*Tippana–Garlapati model (six parameters model)*⁵². It is a degree of freedom model and mathematically stated as:

$$y_2 = \left(A_{12} + B_{11}P_r + C_{11}P_r^2\right)T_r^2 + \left(D_{11} + E_{11}P_r + F_{11}P_r^2\right)$$
(16)

where $A_{11} - F_{11}$ are model constants.

New model. According to solid–liquid phase equilibrium criteria, the fugacity of the solute in the solid phase and liquid phase is equal at equilibrium. The liquid phase is considered as an expanded liquid phase of $ScCO_2$. At equilibrium, the solubility may be expressed as^{53–57}

$$y_2 = \frac{1}{\gamma_2^{\infty}} \frac{f_2^S}{f_2^L} \tag{17}$$

where γ_2^{∞} is drug activity coefficient at infinitesimal dilution in ScCO₂ and f_2^S and f_2^L are fugacities of drug in the solid and ScCO₂ phases, respectively. The f_2^S/f_2^L ratio may be expressed as follows:

$$\ln\left(\frac{f_2^S}{f_L^L}\right) = \frac{\Delta H_2^m}{RT} \left(\frac{T}{T_m} - 1\right) - \int_{T_m}^T \frac{1}{RT^2} \left[\int_{T_m}^T \left[\Delta C_p\right] dT\right] dT$$
(18)

where, ΔC_p is heat capacity difference of the drug in solid phase and that of SCCO₂ phase. The terms that include ΔCp is much smaller than the term that has ΔH_2^{m58} , thus leaving ΔCp term yields a much simpler expression for fugacity ratio as:

$$\ln\left(\frac{f_2^S}{f_L^I}\right) = \frac{\Delta H_2^m}{RT} \left(\frac{T}{T_m} - 1\right) \tag{19}$$

Combining Eq. (19) with Eq. (17) give the expression for the solubility model (Eq. (20)).

$$y_2 = \frac{1}{\gamma_2^{\infty}} \exp\left[\frac{\Delta H_2^m}{RT} \left(\frac{T}{T_m} - 1\right)\right]$$
(20)

In order to use Eq. (20), the appropriate model for γ_2^{∞} is essential.

In this work, the required activity coefficient is obtained from Wilson activity coefficient model⁵⁶ at infinite dilution and it is given by the Eq. (21).

$$\gamma_2^{\infty} = \exp\left[-\ln\left(\lambda_{21}\right) + 1 - \lambda_{12}\right]$$
(21)

where $\lambda_{12} = (V_2/V_1) \exp(-a_{12}/RT)$ and $\lambda_{21} = (V_1/V_2) \exp(-a_{21}/RT)$, V_1 and V_2 are molar volumes of solvent and solute, respectively.

When $\rho_1 = 1/V_1$, the final expression for the infinite dilution activity coefficient is obtained as:

$$\gamma_2^{\infty} = \exp\left[1 + \ln\left(\rho_1 V_2\right) + \frac{a_{21}}{RT} - \rho_1 V_2 \exp\left(\frac{-a_{12}}{RT}\right)\right]$$
(22)

The quantities a_{12} and a_{21} are assumed to be functions of reduced solvent density⁵⁷, and molar volume of the solute is assumed as a constant value. In this work, a_{12} and a_{21} are assumed to have the following form:

$$a_{12} = A(\rho_r)^B \tag{23}$$

$$a_{21} = C(\rho_r)^D \tag{24}$$

Combining Eqs. (22), (23) and (24) with Eq. (20), give the following new explicit solubility model:

$$y_2 = \exp\left[\frac{\Delta H_2^m}{RT} \left(\frac{T}{T_m} - 1\right)\right] / \exp\left[1 + \ln\left(\rho_1 V_2\right) + \frac{A(\rho_r)^B}{RT} - \rho_1 V_2 \exp\left(\frac{-C(\rho_r)^D}{RT}\right)\right]$$
(25)

Equation (25) has four temperature independent adjustable variables namely A,B,C and D.

Equation of state (EoS) model. The solubility of drug i (solute) in ScCO₂ (solvent) is expressed as⁵⁹⁻⁶¹:

$$y_{i} = \frac{P_{i}^{S}\hat{\varphi}_{i}^{S}}{P\hat{\varphi}_{i}^{SeCO_{2}}} \exp\left[\frac{\left(P - P_{i}^{S}\right)V_{S}}{RT}\right]$$
(26)

where P_i^s is the sublimation pressure of the pure solid at system temperature T, P is the system pressure, V_s is the molar volume of the pure solid, R is the universal gas constant. The fugacity coefficient of the pure solute at saturation ($\hat{\varphi}_i^S$) is usually taken to be unity. In this work, the fugacity coefficient of the solute in the supercritical phase $\hat{\varphi}_i^{ScCO_2}$ is calculated using EoS along with KMmr⁵⁷. The expression used for calculation of $\hat{\varphi}_i^{ScCO_2}$ is obtained from the following basic thermodynamic relation⁶⁰:

$$\ln\left(\hat{\varphi}_{i}^{ScCO_{2}}\right) = \frac{1}{RT} \int_{v}^{\infty} \left[\left(\frac{\partial P}{\partial N_{i}}\right)_{T,V,N_{j}} - \frac{RT}{v} \right] dv - \ln Z$$
(27)

The expression for $\hat{\varphi}_i^{ScCO_2}$ is

$$\ln\left(\hat{\varphi}_{i}^{ScCO_{2}}\right) = \ln\left(\frac{v}{v-b}\right) + \left(\frac{2\sum x_{j}b_{ij}-b}{v-b}\right) - \ln\left(Z\right) \\ + \left(\frac{a(2\sum x_{j}b_{ij}-b)}{b^{2}RT^{3/2}}\right) \left[\ln\left(\frac{v+b}{v}\right) - \frac{b}{v+b}\right] \\ \left(3\alpha^{1/2}\left(\sum x_{j}a_{ij}^{2/3}b_{ij}^{1/3}\right)/b^{1/2} - \alpha^{2/3}\left(\sum x_{j}b_{ij}/b^{3/2}\right)\right)/bRT^{3/2}$$

$$(28)$$

where $\alpha = \sum_{i}^{n} \sum_{j}^{n} x_{i} x_{j} a_{ij}^{2/3} b_{ij}^{1/3}$ and the associated mixing rules are:

$$a = \frac{\left(\sum_{i}^{n} \sum_{j}^{n} x_{i} x_{j} a_{ij}^{2/3} b_{ij}^{1/3}\right)^{3/2}}{\left(\sum_{i}^{n} \sum_{j}^{n} x_{i} x_{j} b_{ij}^{1/2}\right)^{1/2}}$$
(29)

$$b = \sum_{i}^{n} \sum_{j}^{n} x_i x_j b_{ij} \tag{30}$$

$$a_{ij} = \left(1 - k_{ij}\right) \sqrt{a_{ii}a_{jj}} \tag{31}$$

$$b_{ij} = \left(1 - l_{ij}\right) \frac{\left(b_{ii}^{1/3} + b_{jj}^{1/3}\right)^3}{8}$$
(32)

The main reason for considering RKEoS is that it has only two adjustable constants k_{ij} and l_{ij} . All the models (density-based, new and RKEoS models) are correlated with the following objective function⁵⁸:

$$OF = \sum_{i=1}^{N} \frac{|y_{2i}^{\exp} - y_{2i}^{calc}|}{y_{2i}^{\exp}}$$
(33)

The regression ability of a model is indicated in terms of an average absolute relative deviation percentage (AARD %).

AARD (%) =
$$(100/N_i) \sum_{i=1}^{N} \frac{|y_{2i}^{exp} - y_{2i}^{cal}|}{y_{2i}^{exp}}$$
 (34)

For the regression, fminsearch (MATLAB 2019a^{*}) algorithm has been used.

Results and discussion

Table 1 shows some physicochemical properties of the used materials. *Empagliflozin* solubility in ScCO₂ is reported at various temperatures (T = 308 to 338 K) and pressures (P = 12 to 27 MPa). Table 2 indicates the solubility data and ScCO₂ density. The reported ScCO₂ density is obtained from the NIST data base. Figure 3 shows the effect of pressure on various isotherms. The cross over region is observed at 16.5 MPa. From Fig. 3, below the cross over region, solubility decreases with increase in temperature, and on the other hand, above the cross over region, the solubility increases with increase in temperature. The EoS model requires critical properties which are computed with standard group contribution methods based on the chemical structure^{62–65}. The summary of the critical properties computed are shown in Table 3. Figure 4 presents the self-consistency of the measured data with MT model.

The density-based models considered in this work have different number of adjustable parameters. These parameters range from three to six numbers. The regression results of all the models are indicated in Tables 4 and 5. The correlating ability of the models is depicted in Figs. 5, 6, 7, 8, 9, 10, 11. From the results, it is clear that all the models are able to correlate the data reasonably well and maximum AARD% is observed to be 10.4%. It is believed that, more parameter models are able to correlate the data more accurately. Sodefian et al., model is able to correlate the data with AARD = 5.84% and Akaike's information criterion (AIC = - 637.59) (more relevant information is presented in the following section). Among density models, Bian et al., model (five parameters model) is able to correlate the data well and corresponding AARD% is 5.1%. Interestingly, Chrastil (three parameters model) and Reformulated Chrastil models (three parameters model) are also able to correlate the data quite well. Further, Chrastil and Reformulated Chrastil models are able to provide the total enthalpy. Whereas, Bartle et al., model parameters are able to provide sublimation enthalpy of the *empagliflozin* drug. From the magnitude difference between the total and sublimation enthalpies, a solvation enthalpy is calculated. These results are reported in Table 6.

A new explicit solubility model based on solid–liquid equilibrium criteria combined with Wilson activity coefficient model corresponding to infinitesimal dilution is derived. The new model has four parameters *A*, *B*, *C* and *D*. While regression, new model parameters are treated as temperature independent and solid molar volume is kept constant. The new model requires melting point, melting enthalpy and molar volume of *empagliflozin* drug, and these values are obtained from literature and group contribution methods. From literature³¹, the melting point of *empagliflozin* drug (426.1 K), molar volume (3.2699×10⁻⁴ m³/mol) and melting enthalpy (60.238 kJ/mol) are calculated based on literature, Immirzi and Perini⁶³ and Jain et al., methods⁶⁶, respectively. The new model makes use of objective function given in Eq. (33). Similarly, RKEoS along KMmr correlations are established with the help of critical properties given in Table 3 (temperature independent correlations). The optimization results of the new solubility and RKEoS models are indicated in Table 5.

In order to examine the ability of models for correlating the solubility data, AIC is applied^{67–70}. When the data number is less than < 40, the corrected AIC (AIC_c) is used.

Temperature (K) ^a	Pressure (MPa) ^a	Density of ScCO ₂ (kg/ m ³) ⁷¹	$y_2 \times 10^4$ (mole fraction)	Experimental standard deviation, $S(\bar{y}) \times (10^4)$	S (equilibrium solubility) (g/L)	Expanded uncertainty of mole fraction (10 ⁴ U)
	12	769	0.0814	0.0021	0.0643	0.0055
	15	817	0.1266	0.0042	0.1060	0.0098
308	18	849	0.1327	0.0010	0.1156	0.0062
508	21	875	0.1411	0.0051	0.1265	0.0118
	24	896	0.1501	0.0063	0.1378	0.0137
	27	914	0.1806	0.0071	0.1692	0.0161
	12	661	0.0706	0.0023	0.0479	0.0052
	15	744	0.1182	0.0031	0.0901	0.0081
210	18	791	0.1515	0.0032	0.1228	0.0091
518	21	824	0.1601	0.0041	0.1353	0.0107
	24	851	0.2040	0.0064	0.1812	0.0151
	27	872	0.2079	0.0093	0.1858	0.0202
	12	509	0.0611	0.0031	0.0319	0.0066
	15	656	0.1044	0.0023	0.0702	0.0062
229	18	725	0.1620	0.0032	0.1203	0.0094
528	21	769	0.1860	0.0042	0.1467	0.0115
	24	802	0.2248	0.0091	0.1849	0.0206
	27	829	0.2260	0.0021	0.1920	0.0107
	12	388	0.0514	0.0023	0.0204	0.0047
220	15	557	0.0928	0.0011	0.0530	0.0047
	18	652	0.2002	0.0101	0.1338	0.0219
556	21	710	0.2266	0.0112	0.1650	0.0242
	24	751	0.2637	0.0103	0.2030	0.0231
	27	783	0.2590	0.0091	0.2079	0.0213

Table 2. Solubility of crystalline empagliflozin in ScCO₂ at various temperatures and pressures. The

experimental standard deviation was obtained by $S(y_k) = \sqrt{\sum_{j=1}^{n} (y_j - \bar{y})^2 \frac{1}{n-1}}$. Expanded uncertainty (U) = $k^* u_{combined}$ and the relative combined standard uncertainty $u_{combined}/y = \sqrt{\sum_{i=1}^{N} (P_i u(x_i)/x_i)^2}$ a Standard uncertainty u are $u(T) = \pm 0.1$ K; $u(p) = \pm 0.1$ MPa. The value of the coverage factor k = 2 was chosen on the basis of the level of confidence of approximately 95 percent.



Figure 3. *Empagliflozin* solubility in ScCO₂ vs. pressure.

					T(K)			
					P _{sub} (Pa) ^e			
Substance	T _c (K)	P _c (Pa)	ω	V ^s ×10 ⁻⁶ (m ³ /mol)	308	318	328	338
Empagliflozin	870.367 ^a	18.7565 ^b	0.479 ^c	184.397 ^d	0.0034	0.0089	0.022	0.0508
CO ₂	304.18	73.8	0.225					

Table 3. Critical and physical properties of *empagliflozin* and CO₂. T_c: critical temperature; P_c: critical pressure; ω : acentric factor; V^s: solid molar volume; T: temperature. ^aEstimated by Fedors method. ^bEstimated by the Joback and Reed method. ^cEstimated by Lee-Kesler vapour pressure relations (the required normal boiling temperature (at 1.0 atm), T_b is estimated with Klincewicz relation, T_c = 50.2–0.16 M + 1.41 Tb, where M is molecular weight). ^dEstimated by Immirzi, A., Perini, B method. ^eEstimated by Lee-Kesler vapour method.



Figure 4. Self-consistency plot based on MT model.

Name of the empirical model	Correlation parameters	AARD%	R ²	R ² _{adj}
Alwi-Garlapati model	$A_1 = -1.8293; B_1 = -14.218; C_1 = 2.8519$	6.58	0.941	0.932
Bartle et al., model	$A_2 = 12.195; B_2 = -5972.3; C_2 = 7.7336 \times 10^{-3}$	10.4	0.922	0.910
Bian et al., model	$A_3 = -0.062205; B_3 = -5.7629 \times 10^{-4}; C_3 = -6230.8; D_3 = 2.9473; E_3 = 4.5582$	5.1	0.951	0.938
Chrastil model	$\kappa = 3.9083; A_4 = -18.97; B_4 = -3674.3$	9.21	0.943	0.934
Garlapati-Madras model	$A_5 = -750.2; B_5 = 852.82; C_5 = 1.0855; D_5 = -7397.7; E_5 = -11.163$	7.09	0.946	0.930
Kumar–Jonstone model	$A_6 = -14.274; B_6 = -0.53652; C_6 = 2.1216$	27.3	0.902	0.892
Mahesh_Garlapati model	$A_7 = -14.266; B_7 = -0.52714; C_7 = 2.0972$	8.14	0.931	0.921
Mendez-Teja model	$A_8 = -7775.4; B_8 = 2.3557; C_8 = 12.694$	9.95	0.924	0.912
Sodefian et al., model	$A_9 = -23.94; B_9 = 1.6043 \times 10^{-3}; C_9 = 2.4939; D_9 = 2.6639 \times 10^{-4}; E_9 = -9.5238 \times 10^{-3}; F_9 = -1.037 \times 10^{-3};$	5.84	0.956	0.940
Reformulated Chrastil model	$\kappa' = 3.8748; A_{10} = -33.58; B_{10} = -2705.8$	9.14	0.943	0.935
Tippana–Garlapati model	$\begin{array}{l} A_{11} = - \ 6.4027 \times 10^{-5}; B_{11} = - \ 1.1813 \times 10^{-5}; C_{11} = 2.0367 \times 10^{-5}; D_{11} = 4.4544 \times 10^{-5}; E_{11} = 3.5989 \times 10^{-5}; F_{11} = - 2.4670 \times 10^{-5} \end{array}$	6.63	0.927	0.924

Table 4. Correlation constants for the exiting empirical models.

Model	Correlation parameters	AARD%	R ²	R ² _{adj}
New model	A = 36,634; B = -0.096039; C = -9673.6; D = -0.16480	7.22	0.949	0.941
RKEoS + Kwak Mansoori mixing rule model	$k_{ij} = 0.32061; l_{ij} = 0.1949$	8.07	0.951	0.946

 Table 5.
 Calculated result for the new model and RKEoS + Kwak-Mansoori mixing rule model.



Figure 5. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines and broken lines are calculated solubilities with Chrastil and Reformulated Chrastil models, respectively.



Figure 6. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines and broken lines are calculated solubilities with KJ and Bartle et al., models, respectively.

$$AIC_{c} = AIC + \frac{2Q(Q+1)}{N-Q-1}$$
(35)

where AIC, N, Q and SSE are N ln (SSE/N) + 2Q, the number of observations, the number of adjustable parameters of the model and the error sum of squares, respectively. According to AIC_c criterion, the best model has the least AIC_c value. Table 7 shows AIC_c values for various models considered in this study. In terms of AIC_c, all the models are able to correlate the data closely. However, Reformulated Chrastil model has AIC_c value (-637.02), hence it is treated as the best model and at the same time, Tippana–Garlapati model has the highest AIC_c value (-621.69), therefore, it is considered as a poorly correlating model. Three parameters models namely Chrastil, Alwi–Garlapati and Mendez–Teja models have AIC_c values -636.95, -635.3 and -635.4, respectively. The new model which has four parameters, indicating comparable performance with the best model (AIC_c value of -637.24).

Conclusion

Solubilities of *empagliflozin* in ScCO₂ at temperatures (T = 308-338 K) and pressures (P = 12-27 MPa) were reported for the first time. The measured solubility in terms of mole faction ranged from 5.14×10^{-6} to 25.9×10^{-6} . The data was successfully correlated with several models, Bian et al., model (AARD = 5.1%) was observed to be the best model in correlating the solubility data. All the models are able correlate the data reasonable. However,



Figure 7. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines and broken lines are calculated solubilities with Alwi-Garlapati and Mahesh-Garlapati models, respectively.



Figure 8. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines and broken lines are calculated solubilities with Bian et al., and Garlapati–Madras models, respectively.

the correlating ability in ascending order of various models in terms of the lowest AIC_c values is as follows: Bian et al., Reformulated Chrastil, Chrastil, new solid–liquid equilibrium, Mendez–Teja, RKEoS + KMmr, Alwi–Garlapati, Sodefian et al., Mahesh–Garlapati, Bartle et al., Tippana–Garlapati models. The new model proposed in this work may be useful for correlating solids solubility in any SCF.



Figure 9. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines and broken lines are calculated solubilities with Tippana–Garlapati and Sodeifian et al., models, respectively.



Figure 10. *Empagliflozin* solubility vs. ScCO₂ density. Solid lines are calculated solubilities with new model.



Figure 11. *Empagliflozin* solubility vs. pressure. Solid lines are calculated solubilities with RKEoS + KM mixing rule.

	Thermodynamic property					
Model	Total entalpy, ΔH _{total} (kJ/mol)	Enthalpy of sublimation, ΔH_{sub} (kJ/mol)	Enthalpy of solvation, ΔH_{sol} (kJ/mol)			
Chrastil model	30.548ª		- 19.105 ^d			
Reformulated Chrastil model	22.496 ^b		– 27.157 ^e			
Bartle et al., model		49.653 ^c (approximate value)				

Table 6. Computed thermodynamic properties. ^aObtained with Chrastil model ^bObtained with Reformulated Chrastil model ^cObtained with Bartle et al. ^dObtained as a result of difference between the ΔH_{sub}^{c} and ΔH_{total}^{a} . ^eObtained as a result between the ΔH_{sub}^{c} and ΔH_{total}^{b} .

Model SSE 10¹¹ RMSE 10⁶ N O AIC AIC

Model	SSE.10 ¹¹	RMSE.10 ⁶	N	Q	AIC	AIC _c
Existing density models						
Alwi-Garlapati model	5.673	1.537	3	24	- 636.5	- 635.30
Bartle et al., model	7.391	1.755	3	24	- 630.15	- 628.95
Bian et al., model	4.338	1.345	5	24	- 638.94	- 635.6
Chrastil model	5.297	1.486	3	24	- 638.15	- 636.95
Garlapati-Madras model	4.828	1.418	5	24	- 636.37	- 633.04
Kumar-Jonstone model	6.537	1.650	3	24	- 633.1	- 631.9
Mahesh_Garlapati model	6.337	1.625	3	24	- 633.84	- 632.64
Mendez-Teja model	5.650	7.51	3	24	- 636.60	- 635.4
Sodefian et al., model	4.222	1.326	6	24	- 637.59	- 632.65
Reformulated Chrastil model	5.280	1.483	3	24	- 638.22	- 637.02
Tippana–Garlapati model	6.659	1.666	6	24	- 627.0	- 621.69
New model						
New solid-liquid equilibrium model	4.635	1.389	4	24	- 639.35	- 637.24
EoS model		-				
RKEoS model + Kwak-Mansoori mixing rule	6.201	1.607	2	24	- 636.36	- 635.79

Table 7. Statistical quantities (SSE, RMSE, AIC and AIC_c) of various models.

Data availability

The datasets generated and/or analysed during the current study are not publicly available due to confidential cases are available from the corresponding author on reasonable request.

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Author contributions

G.S. conceptualization, methodology, validation, investigation, supervision, project administration, writing—review and editing; C.G. methodology, investigation, software, writing—original draft; F.R. investigation, validation, resources; H.N. measurement.

Competing interests

The authors declare no competing interests.

Additional information

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