

Modeling the Phase Equilibria of H₂O – CO₂ mixture with PC-SAFT and tPC-PSAFT equations of state

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Abstract

Water – carbon dioxide binary mixture is important for a number of industrial and environmental applications. Accurate modeling of its thermodynamics properties is a challenging task due to the highly non-ideal intermolecular interactions. In this work, two models based on the Statistical Associating Fluid Theory (SAFT) are used to correlate reliable experimental vapor – liquid equilibria (VLE) and liquid – liquid equilibria (LLE) data in the temperature range 298 – 533 K. CO₂ is modeled as a non-associating or associating component within the Perturbed Chain-SAFT (PC-SAFT) and as a quadrupolar component within the truncated PC-Polar SAFT (tPC-PSAFT). It is shown that PC-SAFT with explicit account of H₂O – CO₂ cross-association and tPC-PSAFT with explicit account of CO₂ quadrupolar interactions are the most accurate from the models examined. Saturated liquid mixture density data are accurately predicted by the two models.

Keywords: Carbon capture and storage, water, carbon dioxide transport, equation of state, polar fluids

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1. Introduction

Water – carbon dioxide ($\text{H}_2\text{O} - \text{CO}_2$) mixture is very important for a number of industrial and environmental applications. Being the leading greenhouse gas, CO_2 capture and sequestration has attracted significant attention by scientists and engineers [1]. The most widely accepted approaches to sequester CO_2 are two: either in the beds of deep oceans [2], or in oil reservoirs where the injected CO_2 can act as enhancer for the recovery of oil (enhanced oil recovery – EOR) in the case of a not depleted well [3]. All these processes have a common point that is the continuous appearance of mixtures of CO_2 with other components, such as hydrocarbons, gases, or H_2O [4, 5]. $\text{H}_2\text{O} - \text{CO}_2$ mixture is also important for CO_2 transport via pipelines, since the flue gas that is transported is not totally dry after the separation processes [6-8]. In addition, in the event of a pipeline rupture, CO_2 immediately mixes with the humidity of the atmosphere. One of the most notable design considerations is that CO_2 is an acid gas and in the presence of H_2O may react to form carbonic acid. The corrosion that stems from the formation of carbonic acid is the main challenge for processes that involve CO_2 , something that in oil and gas industry is referred to as “sweet gas” corrosion [9]. Also, the formation of hydrates is potent, under certain conditions [10].

Thermodynamic properties and in particular phase equilibria of $\text{H}_2\text{O} - \text{CO}_2$ has been studied both experimentally and computationally extensively in the last decades. In most of these approaches, strong intermolecular interactions between H_2O and CO_2 molecules are modeled using an association scheme based on Wertheim’s perturbation theory. Recently, Tsivintzelis et al. [11] modeled the vapor – liquid equilibria (VLE) of the mixture using the cubic plus association (CPA) equation of state (EoS) coupled with two different approaches for the estimation of the cross-association parameters. In the first approach, the arithmetic mean is used for the energy of association and the geometric mean for the volume of association. The second approach uses values from experimental studies that can be found in the literature. They concluded that for CO_2 , cross-association with H_2O should be taken into account but self-association should not be included. Also,

due to the high non-ideality of the system, a non-zero binary interaction parameter is required.

Pappa et al. [12] used the Peng-Robinson (PR) EoS in three different variations to study this mixture. Two of them refer to the mixing rules implemented (van der Waals one-fluid and universal mixing rules) and the third approach takes into account the association interactions, using the CPA-PR EoS. In this case, both H₂O and CO₂ are modeled as 4-associating site molecules. This modeling scheme for CO₂ was used by the same group in other studies as well [13, 14]. Their calculations for the solubility of CO₂ in H₂O yield errors less than 25.5 % for temperatures lower than 373 K, while for higher temperatures the maximum error observed was 27.8 %. They concluded that for temperatures lower than 373 K the three models perform similarly. At higher temperatures, the CPA-PR approach is superior over the other two.

The Statistical Associating Fluid Theory (SAFT) theory and its variations (Perturbed Chain-SAFT, etc.) are suitable approaches to model the H₂O – CO₂ mixture. Ji et al. [15] used the SAFT1-RPM EoS, a SAFT-based electrolyte EoS, in order to model H₂O – CO₂ mixture and the effect of NaCl in it. CO₂ was modeled as a 3-associating site molecule. The parameters of energy and volume of cross-association were fitted to experimental data and shown to be temperature dependent, especially for temperatures lower than 373 K. They argue that the temperature dependency of these parameters accounts implicitly for the polar interactions.

Karakatsani et al. [16] used the tPC-PSAFT EoS and modeled CO₂ as a quadrupolar fluid with two sites available only for cross-association with H₂O. H₂O was accounted as 4-sites associating dipolar molecule [17]. A linear temperature dependent binary interaction parameter was used. The non-linear pressure change with composition was very accurately reproduced.

An application of the polar version of PC-SAFT EoS, namely PCP-SAFT, for the same mixture was performed by Tang and Gross [18]. They accounted explicitly for the quadrupole – quadrupole interactions of CO₂, while H₂O was treated as strongly associating component without polar interactions. They reported very good results for

temperatures lower than 373 K, with a temperature dependent binary interaction parameter.

In a recent work, Nguyen-Huynh et al. [19] used the group contribution version of PPC-SAFT. H₂O was modeled as associating dipolar and CO₂ as cross-associating quadrupolar fluid, respectively. The cross-association energy was fitted to experimental data. They obtained relatively high errors for the composition of the phases in equilibrium, relatively higher compared to earlier studies.

The objective of this work is to evaluate thoroughly the accuracy of two PC-SAFT versions, namely the original PC-SAFT and the truncated PC-Polar SAFT (tPC-PSAFT) for modeling the VLE of H₂O – CO₂ mixture. Highly reliable experimental data over a wide temperature range [20-25] are used for this purpose. H₂O is modeled as a 4-associating site component while CO₂ is modeled as either non-associating, associating with different number of sites or polar component. In all cases, a single temperature independent binary interaction parameter is used. It should be mentioned that tPC-PSAFT is more accurate than PC-SAFT in correlating water – hydrocarbon liquid – liquid equilibria (LLE), because of the explicit inclusion of polar interactions [26].

2. Equations of state

The PC-SAFT EoS was developed by Gross and Sadowski [27], and is an extension of the SAFT EoS [28-31]. The theoretical foundations of these models lie on the first order perturbation theory of Wertheim [32-35]. In perturbation theory, the potential energy of a relative complex molecular fluid is described as the sum of the potential energy of a simple reference fluid and a perturbation or correction term. Usually, the first term is known accurately and the challenging part is the description of the perturbation term. If a suitable perturbation term, as a function of temperature, density or pressure, and composition, is developed, then all the remaining thermodynamic properties can be calculated using standard thermodynamic expressions.

In this respect, PC-SAFT EoS is written as a summation of the residual Helmholtz free energy terms that occur due to different types of molecular interactions in the system

under study. The residual Helmholtz free energy is equal to the Helmholtz free energy minus the Helmholtz free energy of the ideal gas at the same temperature T and density ρ .

The exact formalism of Gross and Sadowski [27, 36] was used in this work. For the case of associating term, the elegant formalism of Michelsen and Hendriks [37] was implemented that results in substantial reduction of computing time.

The PC-PSAFT EoS is an extension of PC-SAFT to account explicitly for polar interactions, developed by Karakatsani and Economou [38]. The truncated version of PC-PSAFT (tPC-PSAFT) is a relatively simple, yet accurate, engineering model. Both PC-PSAFT and tPC-PSAFT use the formalism of Larsen et al. [39] for dipolar and quadrupolar interactions. The full development can be found in the work of Karakatsani and Economou [38].

For a system that consists of associating chains, tPC-PSAFT can be expressed as:

$$\begin{aligned} \frac{A^{res}(T,\rho)}{NRT} &= \frac{a^{res}(T,\rho)}{RT} = \frac{a(T,\rho)}{RT} - \frac{a^{ideal}(T,\rho)}{RT} = \frac{a^{ref}(T,\rho)}{RT} - \frac{a^{disp}(T,\rho)}{RT} \\ &= \frac{a^{hs}(T,\rho)}{RT} + \frac{a^{chain}(T,\rho)}{RT} + \frac{a^{disp}(T,\rho)}{RT} + \frac{a^{assoc}(T,\rho)}{RT} + \frac{a^{polar}(T,\rho)}{RT} \\ &\quad + \frac{a^{ind}(T,\rho)}{RT} \end{aligned} \quad (1)$$

where a is the Helmholtz free energy per mole and the superscripts res, ideal, ref, hs, chain, disp, assoc, polar and ind refer to residual, ideal, reference, hard sphere (monomer reference fluid), chain, dispersion, association, polar and induced polar interactions respectively. Details on the individual terms can be found in the literature [27, 29, 30, 38].

For an associating component, PC-SAFT requires five parameters that are typically fitted to experimental data, in most cases vapor pressure and saturated liquid density from low temperature up to close to the critical point. These parameters are:

- The number of segments, m , in the chain molecule,
- The chain segment diameter, σ_i ,

- The energy of dispersion interactions between segments, ε_i ,
- The association energy between sites of like molecules, $\varepsilon^{A_i B_i}$,
- The volume of association interactions, $\kappa^{A_i B_i}$.

For the case of tPC-PSAFT, two additional parameters account for polar interactions:

- The effective polar segment diameter, σ_p , which is fitted to experimental data, and
- The dipole, μ , or quadrupole, Q , moment of the fluid, which is usually measured experimentally.

The models are extended to mixtures using appropriate mixing and combining rules for the various parameters. Dispersion interaction parameters are calculated from Lorentz – Berthelot combining rules [27]:

$$\sigma_{ij} = \frac{1}{2}(\sigma_i + \sigma_j) \quad (2)$$

$$\varepsilon_{ij} = (1 - k_{ij})\sqrt{\varepsilon_i \varepsilon_j} \quad (3)$$

where k_{ij} is a binary interaction parameter fitted to mixture data.

For the association parameters, two approaches were studied in this work. In the first approach, the cross association energy and volume are calculated according to the combining rules proposed by Gross and Sadowski [36]:

$$\varepsilon^{A_i B_j} = \frac{1}{2}(\varepsilon^{A_i B_i} + \varepsilon^{A_j B_j}) \quad (4)$$

$$\kappa^{A_i B_j} = \sqrt{\kappa^{A_i B_i} \kappa^{A_j B_j}} \left(\frac{\sqrt{\sigma_i \sigma_j}}{\frac{1}{2}(\sigma_i + \sigma_j)} \right) \quad (5)$$

In the second approach, the energy of cross-association is calculated from Eq. 4, while the volume of cross-association is fitted to mixture VLE experimental data.

3. Results and discussion

PC-SAFT and tPC-PSAFT were initially used for pure H₂O and CO₂ to correlate vapor pressure and saturated liquid density data and calculate the model parameters. Different models were tested for the two components. H₂O was modeled as a 4-associating site molecule (4C in the terminology of Huang and Radosz [30]) with two proton donor and two proton acceptor sites per molecule. This is the most commonly used model for H₂O that has been shown by various researchers [40-45] to provide accurate representation of H₂O and aqueous mixture properties. Model parameters were taken from ref. [46]. In the tPC-PSAFT framework, the dipole – dipole interactions were also taken into account and parameters were taken from ref. [47].

CO₂ was modeled with PC-SAFT as non-associating component and as an associating component with 2 (2B), 3 (3B) or 4 (4C) sites per molecule. In the tPC-PSAFT framework, CO₂ was modeled as a non-associating component with quadrupole – quadrupole interactions. For the case of non-associating CO₂, model parameters were taken from ref. [46] for PC-SAFT and ref. [47] for tPC-PSAFT while for the three associating models, parameters were fitted to experimental vapor pressure and saturated liquid density data [48]. The objective function for the minimization is:

$$O.F. = \min \left[\sum_{i=1}^{N_{data}} \left(\left(\frac{P_i^{EoS} - P_i^{NIST}}{P_i^{NIST}} \right)^2 + \left(\frac{\rho_i^{EoS} - \rho_i^{NIST}}{\rho_i^{NIST}} \right)^2 \right) \right] \quad (6)$$

In Table 1, the parameter values for the various models and the accuracy in correlating experimental data are shown. Explicit account of dipole – dipole interactions in H₂O results in decrease of the association energy as one might expect. The accuracy in correlating vapor pressure and liquid density increases. Modeling CO₂ as an associating component results in an improvement in the correlation of the vapor pressure but has almost no effect in the correlation of liquid density. The association energy of CO₂ for all three models is significantly lower than the association energy of H₂O and so association in CO₂ is relatively weak. Explicit account of quadrupolar interactions does

not improve the model accuracy in predicting the pure CO₂ vapor pressure and saturated liquid density.

Accurate modeling of H₂O – CO₂ phase equilibria is a challenging task. Experimental data and *ab initio* calculations have revealed strong intermolecular interactions between unlike molecules [49-51]. The carbon atom in CO₂ is considered to behave as a Lewis type electron acceptor, while the oxygen atom in H₂O acts as a Lewis type electron donor. In the PC-SAFT and tPC-PSAFT formalisms, such interactions are modeled using a solvating (cross-associating) scheme. In Table 2, the various schemes used in this work to model H₂O – CO₂ associating interactions are shown. Furthermore, a temperature independent binary interaction parameter (k_{ij} in eq. 3) was fitted to mixture VLE data in the temperature range 298 – 533 K [20-24, 52, 53]. Finally, in approach 2 the cross association volume, $\kappa^{A B_j}$, was also fitted to experimental data. In Table 2, k_{ij} and $\kappa^{A B_j}$ values for the various models are shown.

The objective function used to calculate the binary interaction parameters is:

$$O.F. = \min \left[\sum_{i=1}^{N_{data}} \left(\left(\frac{K_i^{EoS} - K_i^{expt}}{K_i^{expt}} \right)^2 \right) \right] \quad (7)$$

where $K_i = \frac{y_i}{x_i}$ is calculated either from PC-SAFT or tPC-PSAFT based on a TP-flash calculation. In Table 3, the % AAD in the composition of the liquid and vapor phase from the various models is shown. At the lowest temperature examined (298 K), H₂O – CO₂ exhibit VLE at low pressure, while at pressures higher than 6.4 MPa the system exhibits LLE. In this case, the non-associating model for CO₂ (Case 1) provides accurate correlation of the CO₂ solubility in H₂O (Figure 1a) but poor correlation of the H₂O solubility in CO₂ (Figure 1b). The accuracy improves significantly when H₂O – CO₂ solvation is taken into account (Case 2). By making $\kappa^{A B_j}$ a second fitted parameter in addition to the k_{ij} parameter (Case 2 – Approach 2), the H₂O solubility correlation improves further, but at the expense of CO₂ solubility. Finally, when CO₂ quadrupole – quadrupole interactions are taken into account, model correlation for both phase solubilities significantly deteriorates (Case 6). In Figures 2 and 3, the H₂O solubility in

CO₂ and CO₂ solubility in H₂O at different temperatures with PC-SAFT (Case 2 – Approach 2) and tPC-PSAFT (Case 6) models are shown. At low temperatures and low pressures, PC-SAFT seems to be more accurate. At temperatures higher than 394 K, calculations from the two models are similar. Calculations from the other models are not presented in detail here, but the overall accuracy of the various approaches can be assessed from Table 3. Clearly, explicit account of association and polar interactions in H₂O (Case 7) results in significant reduction of model accuracy.

An obvious way to increase correlation accuracy is through the introduction of additional adjustable mixture parameters. Ji et al. [15] used SAFT1-PRM and proposed third order polynomials for k_{ij} , the cross association energy and volume with a total of eleven fitted parameters. Excellent correlation was obtained but with limited predictive capability.

Pappa et al. [12] treated used the CPA-PR model with k_{ij} and the cross association energy as linearly dependent on temperature. Moreover, the sets of parameters that they reported are different for the temperature ranges 278 – 373 K and 373 – 623 K, thus requiring a total of eight parameters to be regressed in order to achieve the final results. Tsivintzelis et al. [11] used CPA with two adjustable parameters, that are k_{ij} and a parameter related to cross association volume. In essence, approach 2 here is equivalent to that approach. Tsivintzelis et al. [11] results are similar to the results presented here. In Table 4, a summary of different modeling approaches proposed in the literature are presented for the correlation of H₂O – CO₂ VLE using EoS.

In summary, calculations presented in this work reveal that modeling CO₂ within PC-SAFT as a self-associating component does not improve H₂O – CO₂ VLE and LLE correlation. Instead, by treating CO₂ as solvating component, the model accuracy is improved. For the case of tPC-PSAFT, modeling of CO₂ as quadrupolar fluid results in good correlation of mixture data but it is still less accurate than PC-SAFT.

The optimized models developed here were used to predict the density of H₂O – CO₂ mixture. Experimental data [23, 54] and model predictions from PC-SAFT (Case 2) and

tPC-PSAFT (Case 6) at 298.15 K and 332.15 K are presented in Figure 4. The % AAD is 0.6 % and 0.8 % for PC-SAFT and tPC-PSAFT, respectively.

4. Conclusions

PC-SAFT and tPC-PSAFT equations of state were applied for modeling the VLE of H₂O – CO₂ over an extensive range of temperature and pressure. Several association schemes and polar models were examined for CO₂, while two different approaches to evaluation the cross associating parameters were used. Reliable experimental data were used to validate model calculations.

Pure component parameters were fitted to experimental vapor pressure and saturated liquid density data. Inclusion of dipole – dipole interactions for H₂O improves the accuracy of data correlation. Modeling CO₂ as an associating component provides better correlation of pure component data over modeling it as a quadrupolar fluid.

For mixture calculations, explicit account of H₂O – CO₂ solvation effects through an association scheme clearly improves VLE and LLE correlation. A temperature independent binary interaction parameter is sufficient in all cases. Addition of temperature dependence in this parameter provides a marginal improvement in the calculations. By treating the cross-association volume as a second fitted parameter a marginal improvement in the correlation of experimental data is obtained.

The saturated liquid density of the mixture is predicted accurately by both PC-SAFT and tPC-PSAFT models over a broad pressure range. In summary, PC-SAFT with explicit account of H₂O – CO₂ cross-association and with one adjustable parameter (k_{ij}) is the recommended model for the reliable correlation of mixture phase equilibria.

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TABLES

Table 1. PC-SAFT and tPC-PSAFT parameters for H₂O and CO₂ fitted to vapor pressure and saturated liquid density and % AAD between experimental data and model correlation. Temperature range: 275 – 640 K for H₂O and 217 – 301 K for CO₂.

Component	m	σ (Å)	ϵ/k (K)	ϵ^{AB}/k (K)	κ^{AB}	AAD (%)	
						P^{sat}	ρ^{liq}
<i>PC-SAFT</i>							
H ₂ O (4C)	2.1945	2.229	141.66	1804.17	0.2039	1.98	0.83
CO ₂ (inert)	2.6037	2.555	151.04	-	-	0.39	0.88
CO ₂ (2B)	2.2414	2.713	159.00	512.88	0.0283	0.15	0.92
CO ₂ (3B)	2.3056	2.684	156.83	371.15	0.0277	0.13	1.02
CO ₂ (4C)	2.2280	2.731	157.25	307.41	0.0287	0.17	1.18
<i>tPC-PSAFT</i>							
H ₂ O (4C and dipole ^a)	2.8150	2.037	150.71	1575.20	0.3518	0.42	0.49
CO ₂ (quadrupole ^b)	1.9120	2.854	157.97	-	-	0.82	1.04

^a For H₂O, $\mu=1.85D$, $\sigma_p=3.093$ Å and $a=1.49$ Å³

^b For CO₂, $Q=4.30D$ and $\sigma_q=2.974$ Å.

Table 2. Associating models for H₂O and CO₂, binary interaction parameter values and cross-association volumes. In approach 1, the volume of cross association (κ^{AB}) is calculated from eq. (5) while in approach 2 it is fitted to mixture experimental data. Mixture experimental data in the temperature range 298 – 533 K were used to fit k_{ij} and κ^{AB} .

Case #	EoS	H ₂ O model	CO ₂ model	Approach 1 (one fitted parameter)		Approach 2 (two fitted parameters)	
				k_{ij}	κ^{AB}	k_{ij}	κ^{AB}
1	PC-SAFT	4C	Non-associating	-0.0033	-	-	-
2	PC-SAFT	4C	Solvating	0.0376	0.1020*	0.0496	0.1435
3	PC-SAFT	4C	2B	0.0986	0.0749	0.0640	0.0604
4	PC-SAFT	4C	3B	0.1217	0.0741	0.0691	0.0537
5	PC-SAFT	4C	4C	0.1174	0.0754	0.0675	0.0425
6	tPC-PSAFT	4C	Non-associating + quadrupole	-0.0286	-	-	-
7	tPC-PSAFT	4C + dipole	Non-associating + quadrupole	0.0107	-	-	-

* This value was set to $\frac{1}{2}\kappa_{H_2O}$.

Table 3. % AAD in the correlation of the composition of liquid and vapor phases of H₂O – CO₂ mixture in the temperature range 298 – 533 K from the various models.

Case #	%AAD			
	Approach 1		Approach 2	
	x _{CO2}	y _{H2O}	x _{CO2}	y _{H2O}
1	7.5	37.8	-	-
2	9.7	26.8	10.9	21.6
3	10.3	32.3	9.4	22.0
4	12.0	38.9	10.1	21.7
5	12.7	34.5	10.8	21.7
6	15.0	30.0	-	-
7	17.6	63.2	-	-

Table 4. Summary of literature models for H₂O – CO₂ mixture VLE and reported accuracy.

Model	T [K]	No. of adjustable parameters	%AAD	%AAD	Reference
			x _{CO2}	y _{H2O}	
SRK with vdW	243-623	2, for solubility of H ₂ O in CO ₂	93	7.3	[55]
SRK with vdW	243-623	3, for solubility of CO ₂ in H ₂ O	3.5	392	[55]
SRK-Huron Vidal	243-623	7	2.6	7.3	[55]
SRK-Huron Vidal	243-623	5	4	7.5	[55]
SRK-Huron Vidal (Pedersen)	243-623	5	11.6	204	[55]
PR-Henry	278-318	7	2.6	11.7	[25]
PR-Huron Vidal	278-318	3 for every isotherm	4.4	6.4	[25]
UMR-PR	278-373	11	5.6	14	[12]
UMR-PR	373-623	11	12.3	14	[12]
PR	278-373	8	6.2	13.9	[12]
PR	373-623	8	10.1	37.4	[12]
CPA-PR	278-373	8	5.9	9.6	[12]
CPA-PR	373-623	8	7.1	10.3	[12]
CPA (CO ₂ as 4C with solvation)	243-623	3	11.8	20.6	[55]
CPA (CO ₂ as 4C with solvation)	243-623	3	9	16.8	[55]
CPA (CO ₂ as 4C with solvation)	243-623	2	21	28.4	[55]
CPA (CO ₂ as 4C with solvation)	243-623	1	43	20.9	[55]
CPA (CO ₂ as 4C without solvation)	243-623	0 or 1		56	[55]
CPA (CO ₂ as 2B)	298-478	2	16.5	11.9	[11]
CPA (CO ₂ as 3B)	298-478	2	12.6	14.7	[11]
CPA (CO ₂ as 4C)	298-478	2	10.8	15.1	[11]
SAFT-VR	278-318	2	2.2	12	[25]
GC-PPC-SAFT	298-523	2	41.8	50.1	[19]

FIGURES

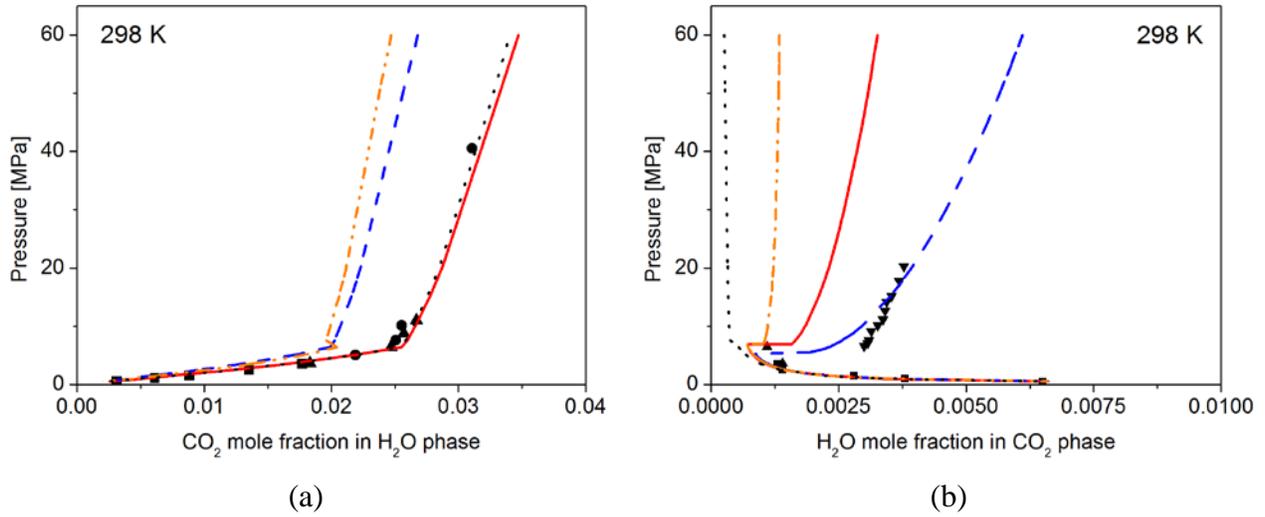


Figure 1. Pressure – composition diagrams for the H₂O - CO₂ phase equilibria at 298 K: (a) Aqueous phase and (b) CO₂ phase. Experimental data (■ Valtz [25], ● Wiebe [56], ▲ Nakayama [21], ▼ King [23]), and PC-SAFT model correlation using the different association and polar schemes (····· Case 1, — Case 2 – Approach 1, - - - Case 2 – Approach 2, - · - · Case 6).

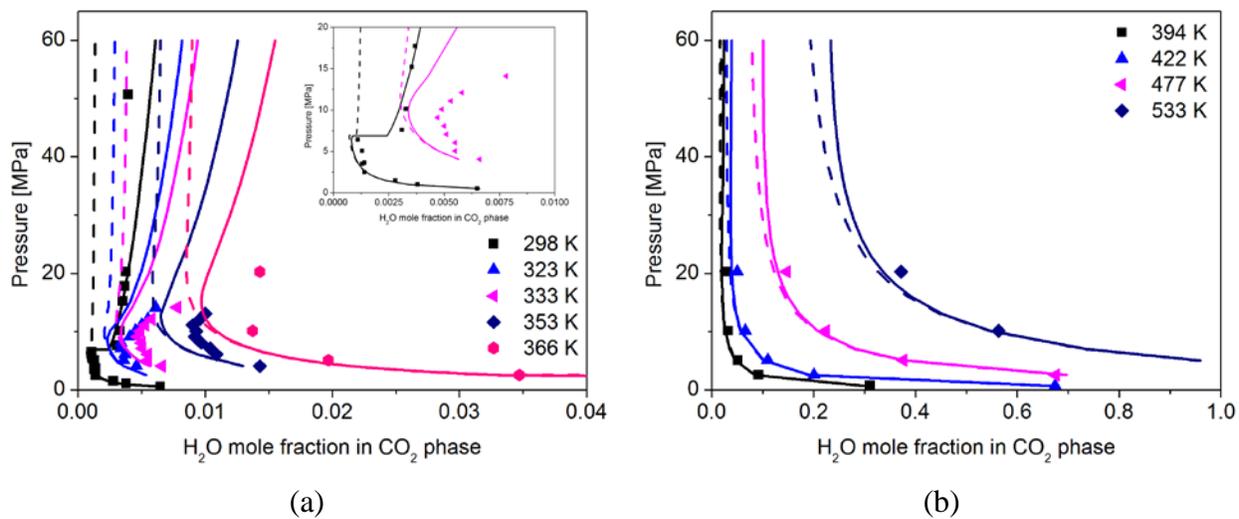


Figure 2. Pressure – composition diagrams for the H₂O solubility in CO₂ at different temperatures. Experimental data (points), PC-SAFT correlation (solid lines; Case 2 - Approach 2), tPC-SAFT correlation (dashed lines; Case 6).

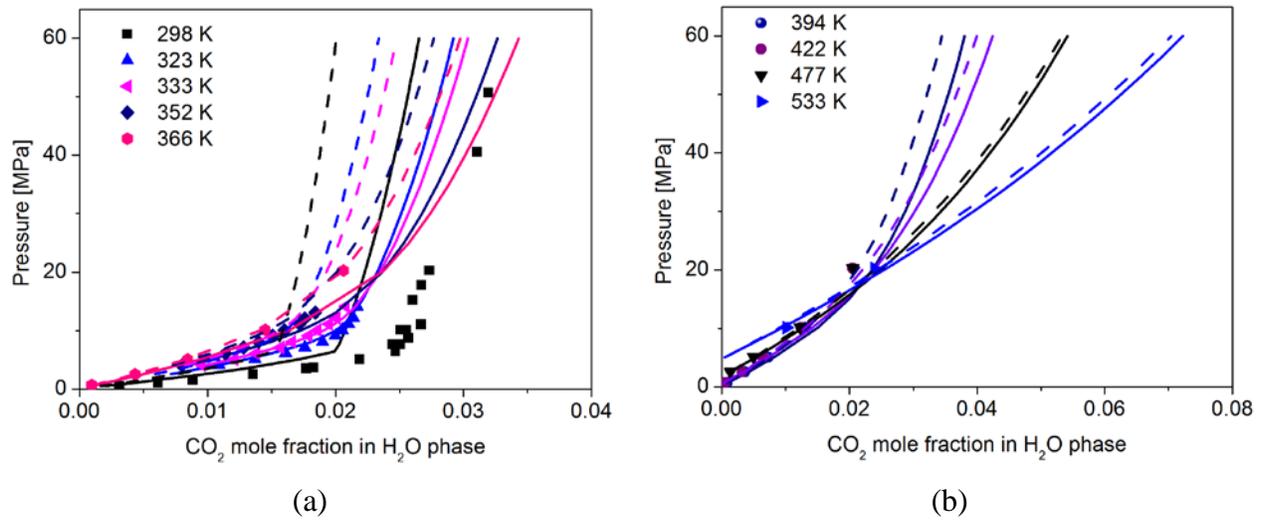


Figure 3. Pressure – composition diagrams for the CO₂ solubility in H₂O at different temperatures. Experimental data (points), PC-SAFT correlation (solid lines; Case 2 - Approach 2), tPC-PSAFT correlation (dashed lines; Case 6).

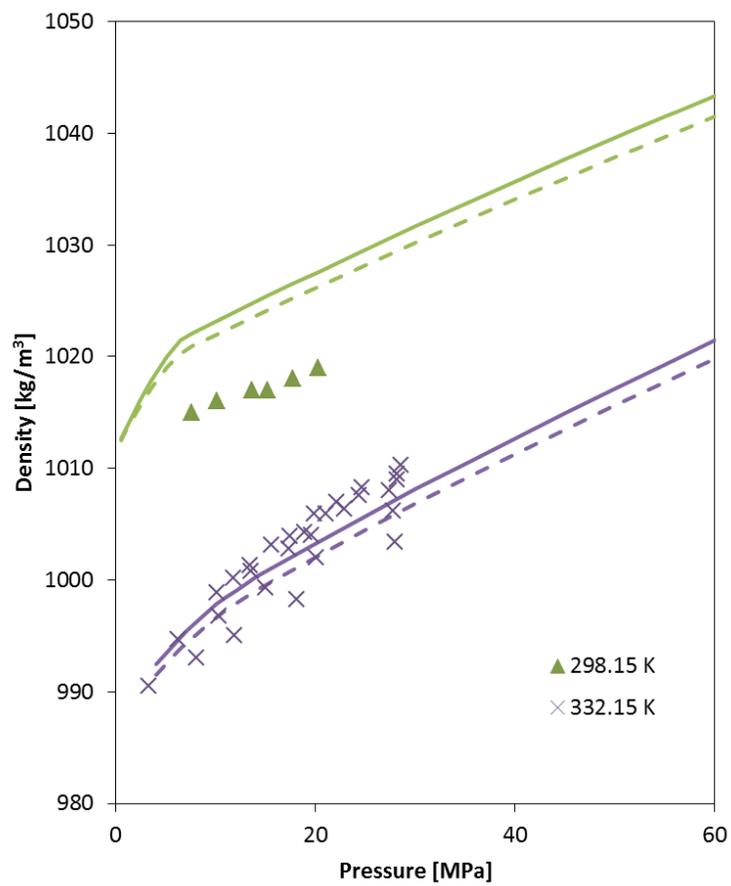


Figure 4. Density of CO₂-saturated H₂O at 298.15 and 332.15 K. Experimental data [23, 53], PC-SAFT prediction (solid lines; Case 2 – Approach 2) and tPC-PSAFT prediction (dashed lines; Case 6).