A Comparative Assessment of Emerging Solvents and Adsorbents for Mitigating CO\textsubscript{2} Emissions From the Industrial Sector by Using Molecular Modeling Tools

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The possibilities offered by molecular modeling tools to obtain relevant data at process conditions, while also gaining molecular insights on the techniques used for CO\textsubscript{2} capture and separation, are presented here using selected case studies. Two different technologies, absorption with amine-based systems and adsorption on porous materials, were explored, using the molecular-based equation of state, soft-Statistical Associating Fluid Theory (SAFT), and Grand Canonical Monte Carlo simulations, respectively. The aqueous monoethanolamine (MEA) system was set as the benchmark for absorption and compared to the performance of 8 alternative amine-based systems, while 16 adsorbents belonging to different families (zeolites, metal–organic frameworks, amorphous silicas, and activated carbons), bare or functionalized with alkylamines, were investigated for the separation of CO\textsubscript{2} by adsorption. In addition to obtaining molecular information on the CO\textsubscript{2} capture process, the models were further used to examine the CO\textsubscript{2} capture performance in terms of cyclic working capacity and energy index as key performance indicators, allowing the identification of promising systems that can improve the current ones to be further evaluated for separation in non-power industries. Results show that for the same total amine mass concentration, non-aqueous amine solvents have a 5–10\% reduction in cyclic working capacity, and a 10–30\% decrease in the energy index compared to their aqueous counterparts due to their lower heat of vaporization and specific heat capacity. In addition, M-MOF-74, NaX, and NaY structures present the best results for adsorption in temperature swing adsorption (TSA) processes. Similar values of energy requirements to those of amine-based systems (2–2.5 MJ kg CO\textsubscript{2}\textsuperscript{−1}) were obtained for some of the adsorbent; however, the disadvantage of the TSA process versus absorption should be considered. These results confirm the reliability of molecular modeling as an attractive and valuable screening tool for CO\textsubscript{2} capture and separation processes.

Keywords: CO\textsubscript{2} capture, absorption, water-free amines, adsorption, metal–organic frameworks, soft-SAFT, molecular simulations
INTRODUCTION

Since the nineteenth century, the planet’s average temperature has risen \( \sim 0.9^\circ \text{C} \), a change driven largely by the increased emissions into the atmosphere of anthropogenic carbon dioxide and other greenhouse gases (GHGs; Santer et al., 2003; Sreedhar et al., 2017), as a consequence of the energy intensive and industrialized society in which we live. Among the different human activities, industry accounts today for \( \sim 40\% \) of the global energy demand and 25% of the total \( \text{CO}_2 \) emissions, the second largest source of emissions after the power sector (International Energy Agency [IEA], 2019). Steel and cement are the two highest-emitting industry subsectors, accounting for more than 10% of the total direct global \( \text{CO}_2 \) emissions: over 2.1 Gt of \( \text{CO}_2 \) per year from each cement and iron/steel industries in the past few years (the chemical subsector is the third largest emitter with around 1.1 Gt/year; International Energy Agency [IEA], 2018).

Although efforts for shifting toward renewable energy sources and more energy-efficient systems are already in place, fossil fuels are likely to remain an important energy source, in the near and medium term (Le Quéré et al., 2019), due to their availability and efficiency. Hence, the \( \text{CO}_2 \) level into the atmosphere, which has already drastically surpassed the safety boundaries (above 400 ppm; Steffen et al., 2015) will continue to increase in a business as usual scenario. Therefore, controlling such emissions is already a critical environmental issue, with scientists and regulators taking actions toward reducing GHG emissions and mitigating climate change (Intergovernmental Panel on Climate Change [IPCC], 2014; National Oceanic and Atmospheric Administration [NOAA], 2019). Significant work and effort have been employed in recent years for the development of \( \text{CO}_2 \) capture technologies to limit the degree of future climate change (Asif et al., 2018) while using fossil fuels. One of the most important alternatives to mitigate anthropogenic \( \text{CO}_2 \) emissions in the short–medium term is by carbon capture, utilization, and storage (CCUS); Wang et al., 2012). Nowadays, 16 large-scale CCUS applications at industrial facilities are capturing more than 30 million tons of \( \text{CO}_2 \) emissions each year from steel production, natural gas processing, and hydrogen and fertilizer (ammonia) production (International Energy Agency [IEA], 2019). Nevertheless, the IEA Clean Technology Scenario, which sets out a pathway consistent with the Paris Agreement climate ambition, mentions that CCUS can contribute with 20% of the emissions reductions needed across the industry sector (i.e., more than 28 Gt of \( \text{CO}_2 \) captured in the period to 2060). A further 31 Gt is expected to be captured from fuel transformation and 56 Gt from the power sector (International Energy Agency [IEA], 2018, 2019). Moreover, the letter “U” appearing in CCUS stands for utilization, as the captured \( \text{CO}_2 \) can also be used in several industrial sectors and products, such as the carbonation of beverages, food preservation, supercritical \( \text{CO}_2 \) extraction processes, enhanced oil recovery (EOR), synthesis of materials for different applications and water treatment, as well as the building block or raw source for other chemicals (Vega, 2010), making the CCUS an alternative, attractive complementary route to \( \text{CO}_2 \) capture and storage.

Several technologies have been explored for \( \text{CO}_2 \) capture and separation (Abanades et al., 2005; MacDowell et al., 2010). Among them, absorption with aqueous solutions of amines is technically proven and considered the most industrially feasible technology for carbon capture at a large scale (Asif et al., 2018; Vega et al., 2019). Monoethanolamine (MEA) is the most used alkanolamine molecule in \( \text{CO}_2 \) capture processes due to its high affinity and reactivity with \( \text{CO}_2 \) and relatively low costs of production. However, an important barrier to the full-scale deployment of any aqueous amine solvents is the high energy required for solvent regeneration, primarily related to the sensible heat of the cosolvent, water (Oyenekan and Rochelle, 2007; Svendsen et al., 2011; Budzianowski, 2015; Zhang et al., 2015; Li et al., 2016). The reduction in this energy penalty can be accomplished through several strategies such as process optimization, the use of different amines, and/or alternative solvents with lower regeneration energy requirements, the use of different techniques such as adsorption on microporous solid materials, among others (Huck et al., 2014; Budzianowski, 2016; Bernhardsen and Knuutila, 2017; Heldebrant et al., 2017). Although alternative absorbents (e.g., water-free and water-leak solvents) and adsorbent (e.g., metal–organic frameworks, MOFs) are quite promising, many of them are still in early stages of development, lacking in most cases required data for rigorous modeling, or large-scale projections with the current state of available modeling tools (Prats et al., 2017; Alkhatib et al., 2019). In this regard, with the improvement in computing power and new algorithms, computational methods have become an emblematic complementary tool to experiments (Smit and Maesen, 2008; Garcia et al., 2017). Appropriate and reliable thermodynamic models that can reproduce the physicochemical behavior of the systems, combined with simple engineering calculations, can provide an efficient framework to assess and screen the performance of these promising materials in terms of key performance indicators (KPIs) such as cyclic working capacities and energy index of regeneration (Bahamon and Vega, 2016; Alkhatib et al., 2019) and determine potential systems to be further explored in pilot and large facilities (Bai et al., 2015).

Herein, we present some examples concerning the use of molecular modeling techniques based on statistical mechanics (molecular-based equations of state and molecular simulations) for the modeling of alternative amine-based absorbents and porous solid adsorbents for \( \text{CO}_2 \) capture and separation. A list of the systems evaluated is presented in Table 1, together with the \( \text{CO}_2 \) concentration explored and the operating conditions, representative of \( \text{CO}_2 \) emissions from different industrial sectors. Accurate and transferable models able to characterize the absorption and adsorption processes are used as a step forward in developing predictive models for these systems, aiming to find a balance between accuracy, transferability, and simplicity. Such treatments have been successfully implemented by our group and others in previous studies (Vega and Bahamon, 2016; Lloret et al., 2017; Bahamon et al., 2018; Pereira and Vega, 2018; Alkhatib et al., 2020). This work serves as a bridge between the fundamental knowledge on the molecular behavior of the substances/materials and how this is translated into their performance at industrial conditions. We illustrate how
molecular-based equations of state and molecular simulations can be used as a robust tool for a double purpose: (1) to understand the underlying phenomena of CO2 capture in absorbent solutions and adsorbent materials and (2) to complement experimental data for process design at industrial conditions of relevance for CO2 capture. In Figure 1, we show a schematic of the followed approach, including the length scales covered by the modeling tools used in this work.

**METHODOLOGY**

The first step into applying molecular modeling tools to experimental systems is to propose and validate the models with available experimental data. Then, a further step includes data generation and data collection for novel systems obtained from these models at typical conditions used in industrial processes. Within this approach, it is possible to evaluate the theoretical limits of the energy requirements to capture CO2 under different specific conditions.

In this section, we present a brief introduction of the two different molecular methods used in this work, as well as the main information required and used for developing the transferable models. Absorption and adsorption results are presented for binary gas mixtures containing 25, 15, and 5% CO2 (see Table 1), in order to cover most of the possible typical compositions encountered in industrial processes, including but not limited to the steel and cement industries (Zhang et al., 2014). In all calculations, it was assumed that nitrogen accounted as the surplus of the composition and that the binary mixture was captured at a temperature of 313 K (i.e., 40°C) and atmospheric pressure. It is well known that flue gas contains roughly ∼5% H2O and trace amounts of impurities such as SO2 (Sumida et al., 2012). However, impurities effect is out of the scope of this work (for contributions on this subject, the reader is referred to Bahamon et al., 2018; Alkhatib et al., 2019 and references therein). Moreover, we constrained the range of operating temperatures, opening the possibility of using sources of energy at low temperatures (e.g., low-pressure steam for heating). A condensing steam was assumed to be available at 413 K, with a minimum approach temperature of 10 K.

**Table 1** Summary of the systems evaluated in this work for CO2 capture, together with the conditions at which they were evaluated.

| CO2 capture technology | Systems evaluated | Conditions |
|------------------------|-------------------|------------|
| Absorption with amine-based solvents | MEA + H2O | Inlet CO2 composition: 5% CO2 |
| | DEA + H2O | 15% CO2 |
| | AMP + H2O | 25% CO2 |
| | MEA + EGME | Absorption (rich): P_total = 100 kPa; T = 313 K |
| | MEA + TEG | Desorption (lean): P = 200 kPa; T_max = 403 K |
| | DEA + TEG | Inlet composition: 0.01% CO2/99.99% N2 |
| | AMP + TEG | 5% CO2/95% N2 |
| | AMP + EG | 15% CO2/85% N2 |
| | AMP + DEG | P_total = 100 kPa; T = 313 K |
| | | Adsorption: P = 110 kPa; T_max = 403 K |
| Adsorption with solid porous materials | Zeolite NaX, Zeolite LTA-4A, Mg-MOF-74 | Inlet composition: 0.01% CO2/99.99% N2 |
| | | 5% CO2/95% N2 |
| | | 15% CO2/85% N2 |
| | Amino-funct. Zn-MOF-74 | P_total = 100 kPa; T = 313 K |
| | Activated carbon from date seeds (AC-DS) | Desorption: P = 110 kPa; T_max = 403 K |

**Statistical Associating Fluid Theory**

The Statistical Associating Fluid Theory (SAFT) proposed by Chapman et al. (1989, 1990) is possibly the most well-known Equation of State (EoS) based on statistical mechanics used today for associating and chain molecules in academia and industry. The equation contains key physical information of the fluids. Its parameters are both physically meaningful and transferable, making SAFT a powerful tool for engineering predictions, especially in some applications for which other classical EoSs fail. The success of the equation has been evidenced by the number of published works since its conception (Tan et al., 2008). Different versions of SAFT have been developed over the years. The most popular ones include the original SAFT and simplified SAFT (Chapman et al., 1989; Huang and Radosz, 1990), the soft-SAFT equation (Blas and Vega, 1997; Pámies and Vega, 2001), the SAFT-VR equation (Gil-Villegas et al., 1997), PC-SAFT (Gross et al., 2001), and SAFT-γ-Mie (Lafitte et al., 2013).

Statistical Associating Fluid Theory equations are written in terms of the residual Helmholtz energy $a^{res}$, where each term in the equation represents different microscopic contributions to the total free energy of the fluid. Results obtained from the soft-SAFT equation derived by Blas and Vega (1997, 1998) are presented and discussed in this contribution. The general expression for soft-SAFT is:

$$a^{res} = a^{ref} + a^{chain} + a^{assoc} + a^{polar}$$  (1)

where the superscripts ref, chain, and assoc refer to the contributions from the Lennard–Jones (LJ) reference monomers (Johnson et al., 1992), the formation of the chain, and the association interactions (Wertheim, 1984a,b, 1986a,b). An additional term, $a^{polar}$, can be added to explicitly account for polar interactions arising from dipole or quadrupole moments (Gubbins and Twu, 1978; Twu and Gubbins, 1978; Jog et al., 2001), respectively.

A molecular model for each compound should be proposed within the SAFT framework in order to apply it to experimental systems. This is usually a coarse-grain model where the most relevant features of the molecular structure are captured with a simplified yet reliable manner. In soft-SAFT, substances are treated as spherical LJ (12–6) segments or LJ chains characterized by a specific set of molecular parameters, accounting for the different interactions governing the behavior of the fluid. With this approach, an associating chain molecule is characterized by...
five molecular parameters: the diameter of the segments making the chains \(\sigma\), the energy of interaction between segments \(\varepsilon\), the chain length \(m\), the association volume \(\kappa_{\text{HB}}\), and the association energy \(\varepsilon_{\text{HB}}\). In the case of polar fluids, two additional parameters are needed, being the dipole/quadrupole moment \(\mu/Q\), and the fraction of polar segments in the molecules \(x_p\), which are fixed \textit{a priori} rather than regressed to preserve the robustness of the model (Alkhatib et al., 2020).

The molecular parameters are usually regressed simultaneously using experimental data of liquid density and vapor pressure of the pure substance and are further transferred for the description of other properties of the same fluid (such as heat capacities or speed of sound) as well as multicomponent mixtures. During fitting, it is possible to obtain several sets of parameters providing an accurate description of the data. Hence, apart from the numerical optimization, a critical analysis of the values of the parameters and their physical meaning is required to make them transferable and with extrapolative power (Ferreira et al., 2019). Furthermore, it is possible to extend the equation to calculate additional properties by coupling it to other theories. For instance, treating inhomogeneous systems with the Density Gradient Theory (DGT; Duque et al., 2004; Vilaseca et al., 2010; Vilaseca and Vega, 2011) and/or calculating transport properties with the Free-Volume Theory (FVT; Allal et al., 2001; Llovell et al., 2013). For more details on the implementation of soft-SAFT and other theories, the reader is referred to the original papers (Blas and Vega, 1997; Duque et al., 2004; Llovell et al., 2013; Alkhatib et al., 2020).

Molecular Models Used for Absorption With Alternative Amine-Based Solvents

Although a wide variety of aqueous amine solutions has been studied for the \(\text{CO}_2\) absorption process (Kohl and Nielsen, 1997; Rochelle, 2009; Liang et al., 2015; Puxty and Maeder, 2016), there is still work needed to obtain the best molecules or blends that can overcome the current limitations of the MEA + water system aforementioned. Therefore, absorption with alternative amine-based solvents have been studied in this work using soft-SAFT. Systems assessed include MEA as the benchmark, as well as diethanolamine (DEA), and amino-methyl propanol (AMP). Cosolvents besides water, such as ethylene glycol (EG), diethylene glycol (DEG), triethylene glycol (TEG), and ethylene glycol monomethyl ether (EGME), were also evaluated, effectively forming hybrid chemical-physical solvents when mixed with amines.

Following our previous work (Vega et al., 2009), water was modeled as an LJ sphere with four associating sites, two representing the negative charges lone pair of electrons on the oxygen atom and two representing the positive charges of the hydrogen atoms, while \(\text{CO}_2\) was modeled as a LJ chain with the quadrupolar moment placed in one of the groups (Dias et al., 2006). MEA, DEA, and AMP were modeled as homonuclear
LJ chains, in which their multifunctional groups (amines and hydroxyls) were mimicked with a specific number of association sites for each of them. For details on the models and the allowed interactions, the reader is referred to published works (Lloret et al., 2017; Pereira and Vega, 2018; Pereira et al., 2018; Alkhatib et al., 2019). Conversely, EG, DEG, and TEG were modeled as LJ chainlike molecules with two associating sites of type OH representing the two hydroxyl groups placed at the molecule tails (Pedrosa et al., 2005).

**Figure 2** shows the soft-SAFT models and molecular structures used for CO₂ and H₂O, as well as MEA and EG as representatives of the amines and glycols studied. The large circles on the sketch of the soft-SAFT models stand for the LJ groups making the molecules, while the small circles represent the associating sites, which depend on the specific molecules and type of interactions. In the case of binary mixtures of water and amines, cross-association between the water molecule and the functional groups (NH₂, NH, OH) of each amine molecule were explicitly considered in the calculations (Alkhatib et al., 2019).

In order to make the approach as predictive as possible, all the molecular models and parameters used for the soft-SAFT calculations presented in this work were transferred from previous works. Hence, the molecular parameters of the amines were taken from Lloret et al. (2017), Pereira and Vega (2018), Alkhatib et al. (2019), and Pereira et al. (2018); the molecular parameters for water were transferred from Vega et al. (2009), while the parameters for CO₂ were taken from Dias et al. (2006). The glycol parameters were taken from Pedrosa et al. (2005).

Depending on the type of amine and the type of cosolvent involved in the absorption process, there are different reaction mechanisms governing the absorption of CO₂. The overall reactions can proceed by the formation of carbamate and bicarbonate (Caplow, 1968; Dancwerts, 1978). Note that other thermodynamic models based on classical EoSs, or on activity coefficients, have proved their ability in modeling CO₂ absorption in traditional aqueous amines. However, these models require a considerable number of empirical parameters fitted to experimental data (Al-Rashed and Ali, 2012; Suleman et al., 2015) and the *a priori* knowledge of the reaction equilibrium constants of all relevant ionic species. Following our previous work (Pereira et al., 2018; Alkhatib et al., 2019), the formation of reaction products from the main reactions in the absorption process was implicitly accounted for when using soft-SAFT by the formation of CO₂-amine physical aggregates connected through strong and highly directional association sites. The number of these association sites is dictated by the stoichiometry of the reaction between the CO₂ and the amine. For the specific association values, the reader is referred to the previously mentioned studies (Pereira et al., 2018; Alkhatib et al., 2019). In this work, we are interested in equilibrium properties; however, examining the reaction mechanism governing the formation of carbamate is essential for understanding the effect on the absorption of CO₂ in amines.

**Molecular Simulations—Grand Canonical Monte Carlo**

Molecular simulations comprise a series of techniques based on statistical mechanics, which are capable of predicting phase equilibria and dynamic and structural properties (Allen and Tildesley, 2017). In this regard, statistical mechanics provides a link between the properties of interacting molecules and the thermodynamics of their bulk phases, by introducing...
procedures for averaging over a multitude of possible molecular configurations. The fundamental information required to perform molecular simulations are molecular interactions, which are usually described by a forcefield consisting of bonded (e.g., stretching, bending, and torsional interactions), and non-bonded terms [van der Waals (vdW) and electrostatic interactions] as (Smit and Maesen, 2008):

\[ U = U_{\text{bonded}} + U_{\text{non-bonded}} = [U_{\text{stretching}} + U_{\text{bending}} + U_{\text{torsional}} + \cdots] + [U_{\text{vdW}} + U_{\text{elect}}] \]  

(2)

where \( U \) refers to the total potential energy. Non-bonded interactions are usually considered with LJ potential for the vDW interactions, in a similar way as in the soft-SAFT model (i.e., \( \sigma \) and \( \epsilon \) are comparable), and Coulombic interactions are used for electrostatic interactions.

The two main molecular simulation approaches are Monte Carlo (MC) and Molecular Dynamics (MD; Allen and Tildesley, 2017). MC methods are extensively used for complex molecules calculations, in systems where the temporal evolution is not required, as per calculation of equilibrium properties or for systems for which MD simulations are not affordable due to time scale limitations. Thus, a suitable tool for simulating phase equilibrium systems in confined media, such as adsorption phenomena, is the Grand Canonical Monte Carlo (GCMC) method, which has been extensively used to calculate thermophysical properties in porous materials (Deng et al., 2004; Frost et al., 2006; Herdes et al., 2007; Babarao et al., 2011).

In GCMC, the temperature, volume, and chemical potential of the system are fixed. GCMC allows exchanging molecules between the simulation box containing the porous material and an imaginary ideal gas reservoir, by using a fixed number of MC movements (i.e., insertions, deletions, rotations, and translations; Frenkel and Smit, 2002). The procedure is performed over a large number of steps, and then, the adsorption isotherm can be constructed by calculating the average of the number of molecules adsorbed (after equilibration) for every single pressure (chemical potential) point. For a more comprehensive explanation of this and other molecular simulation methods, the reader is referred to excellent textbooks in the field (Frenkel and Smit, 2002; Allen and Tildesley, 2017).

### Molecular Models Used for Adsorption on Porous Solid Materials

One of the advantages of using adsorption versus absorption for \( \text{CO}_2 \) capture is that compared to the solvents used in the absorption approach, solid adsorbents materials are not volatile and generally do not pose health hazards (Zanco et al., 2018). However, such materials must present high mechanical and thermal stability and easiness of regeneration and availability, ensuring constant high performance during cyclic operation. Additional characteristics desirable for industrially scalable processes include good behavior under humid/impurity conditions and relatively low cost, to name a few (Plaza et al., 2017).

Zeolites and activated carbons have been traditionally used for gas adsorption and separation. These adsorbents can be tuned to the process needs, providing great opportunities for \( \text{CO}_2 \) capture (Prats et al., 2017; Bahamon et al., 2020). Moreover, MOFs have been one of the fastest growing fields in chemistry and materials science, with a vast number of synthesized structures and with publications increasing year by year (Li et al., 2011). Therefore, in this work, we have used GCMC simulations to study the adsorption behavior of \( \text{CO}_2/\text{N}_2 \) mixtures on different families of solid materials. Structures evaluated include zeolites (e.g., NaX, NaY, LTA-4A, and MFI), MOFs (e.g., IRMOF-1, MOF-177, M-MOF-74, CuBTC, and UTSA-16), amorphous silica structures (e.g., MCM-41), carbonaceous materials [e.g., zeolite-templated carbon (FAU-ZTC)], activated carbons (AC-DS; Bahamon et al., 2020), and amino-functionalized structures (see Table 1). The molecular models for the adsorbents were taken from X-ray diffraction structures (when crystals; Olson, 1995; Yaghi et al., 2003; Xiang et al., 2012; Queen et al., 2014) or by representations reported in the literature (Lopez-Aranguren et al., 2015; Bahamon and Vega, 2017). Force fields used for each structure can be found elsewhere (Jorgensen and Tirado-Rives, 1988; Builes et al., 2011; Lopez-Aranguren et al., 2015; Martin-Calvo et al., 2015; Bahamon and Vega, 2016; Prats et al., 2017).

All structures were constructed with simulation box dimensions of at least \( \sim 125 \) Å in each direction, with frameworks treated as rigid structures (Bahamon and Vega, 2016). Simulations were performed by means of LAMMPS (Plimpton, 1995) and Materials Studio software (Dassault Systemes BIOVIA, 2015). For the simulations at each pressure (chemical potential) condition, at least \( 1.0 \times 10^6 \) MC moves were performed for data collection after equilibration. The total energy of the system was calculated as the sum of the adsorbate-adsorbate and the adsorbate-adsorbent interaction energies, modeled as a combination of LJ 12-6 and Coulomb potentials. \( \text{CO}_2 \) molecules—as well as \( \text{N}_2 \)—were modeled using the TraPPE force field (Potoff and Siepmann, 2001). A cutoff radius of 12.5 Å was applied to the LJ interactions, while the long-range electrostatic interactions were calculated by using Ewald summation, and Lorentz-Berthelot combining rules were used to calculate adsorbate/framework and the LJ crossed parameters.

### RESULTS AND DISCUSSION

#### Alternative Amine-Based Solvents for \( \text{CO}_2 \) Capture

Reliable and consistent molecular models were previously developed and validated with pure and binary mixture phase equilibrium data (Pedrosa et al., 2005; Lloret et al., 2017; Pereira and Vega, 2018; Pereira et al., 2018; Alkhatib et al., 2019), providing an accurate prediction with relative overall small deviations with respect to experimental data. Error measurements such as absolute average deviation (AAD) are commonly used to quantify the agreement [although coefficient of determination (\( R^2 \)), index of deviation and mean absolute deviation (MAD) are also reported] (Alkhatib et al., 2019). Models used for the studied amines show% AAD ranging from 2 to 5% for the vapor pressure and an accurate description of the phase behavior of the aqueous solution and \( \text{CO}_2 \) solubility.
The present approach an attractive method for comparison and as a behavior over a broad range of conditions (glycols and glymes is very scarce or inexistent, we have taken (Alkhutib et al., 2019). Additional details can be found in the respective studies. Moreover, since the available experimental data for amines with physical solvents systems studied such as glycols and glymes is very scarce or inexistent, we have taken advantage of the molecular nature of soft-SAFT to predict the behavior over a broad range of conditions \((T, P, i)\). This makes the present approach an attractive method for comparison and as a tool to search for the best operation conditions.

Solubilities of \(CO_2\), \(\alpha_{CO_2}\), in the different amine-based solvents at a fixed amine concentration (i.e., 30 wt%) were calculated from soft-SAFT and are presented as isotherms at 313 and 393 K in Figure 3. If one focuses, for instance, in the absorption at a \(CO_2\) partial pressure of 25 kPa (i.e., 25% \(CO_2\) in the mixture inlet stream at 313 K), it can be seen that higher absorption capacities than MEA + water are achievable with AMP (either with water, EG, DEG, or TEG as solvent). However, if the concentration is decreased to 5% \(CO_2\) (i.e., 5 kPa), a different behavior is observed: in this case, only the aqueous solution with AMP shows higher capacity than MEA + water. In addition, most of the systems display a very low loading at high temperatures, which in principle implies that higher or, at least similar, cyclic capacities to the ones obtained with MEA could be achieved at lower lean temperatures. Moreover, it can be seen that MEA is more prone to change its absorptive behavior with changes in \(CO_2\) concentration than changes in temperature, as can be observed in Figure 3B. DEA and AMP molecules show a more sustained absorption capacity with partial pressure due to the mechanism of \(CO_2\) absorption in these systems, as the aqueous MEA solvent moves from carbamate to bicarbohydrate formation, depending on the amount of \(CO_2\) in the raw flue gas, while the remaining solvents, even with physical cosolvents, absorb \(CO_2\) mainly through the formation of carbamate. Although absorption of \(CO_2\) does not seem to be as efficient with non-aqueous solvents as in aqueous amines in terms of absorption capacity, as also observed by Park et al. (2005), this actually results in lower regeneration temperatures due to the lower enthalpy of vaporization and heat capacity of physical cosolvents compared to water, leading to potential significant reductions in the energy of regeneration (Rivas and Prausnitz, 1979; Hamborg et al., 2011; Zhang et al., 2018). Such a trade-off between solvent absorption capacity and energy of regeneration can be fine-tuned through process optimization.

Toward assessing the performance of solvents for the absorption of \(CO_2\), the absorption cyclic capacity and the regeneration energy index have been recognized as one of the most influential KPIs (Pereira and Vega, 2018) with direct implications on the overall technical and economic feasibility of absorption systems. Hence, for a proper and fair comparison between the different alternative amine-based systems evaluated in this work, we have calculated the cyclic working capacity (in kg \(CO_2\) m\(^{-3}\) sln) and the energy index (in MJ kg \(CO_2^{-1}\)). The cyclic working capacity, \(\Delta \alpha_{CO_2}\), is related to the amount of \(CO_2\) that is effectively separated from the feed gas, defined as the difference between the absorbed amount (solubility) at the feed gas partial pressure and temperature conditions (rich), and the absorbed amount at the operating regeneration conditions (lean):

\[
\Delta \alpha_{CO_2} (T, P) = \alpha_{CO_2, Rich} (T, P_{CO_2})_{Absorber} - \alpha_{CO_2, Lean} (T, P_{CO_2})_{Stripper}
\]  

The amount of \(CO_2\) at both rich and lean conditions were directly calculated from soft-SAFT for the VLE of \(CO_2\) + amine + cosolvent at the required conditions of temperature and pressure. The energy index is defined as the required energy per kilogram of \(CO_2\) desorbed from the rich condition to the lean condition (i.e., the cyclic working capacity). It includes the sensible heat \(Q_{sens}\), the latent heat (for highly volatile systems) \(Q_{vap}\), and the absorption heat \(Q_{abs}\).

\[
E_{index} = \frac{Q_{sens} + Q_{vap} + Q_{abs}}{\Delta n_{CO_2}}
\]

The energy index and its individual contributions were calculated using the shortcut method of Kim et al. (2015), with all required thermophysical properties of pure components, such as heats of vaporization, heat capacities, heats of absorption, and concentrations of \(CO_2\) for the mass balance around the stripping column, obtained from soft-SAFT. Similar comparisons can be found in the literature (Kim et al., 2015; Zhang et al., 2015; Budzianowski, 2016; Li et al., 2016; Bernhardsen and Knuttila, 2017). Nevertheless, note that the comparison presented here is performed in volumetric units (cubic meter of solution, containing 30 wt% amine, and 70 wt% cosolvent), which implies the same volume of absorber equipment for all the different studied systems (Bahamon and Vega, 2016).

The evolution of the lean loading conditions (desorption) was evaluated assuming a stripper column operating at a total pressure of 200 kPa (Pereira and Vega, 2018). Figure 4 depicts the results for the three different \(CO_2\) compositions: 25, 15, and 5%. Values were compared with MEA as a benchmark, creating four quadrants in the chart: the best option(s) must show lower energy index with higher cyclic working capacities and hence must appear in the lower right quadrant. Moreover, the size of the sphere in Figure 4 represents the required temperature to achieve such conditions, since the desorption/lean temperature depends on the specific solvent: the larger the sphere size, the higher the temperature used for regeneration, evaluated with the same shortcut method and typical solvent regeneration process configuration.

Note that, depending on the operating conditions, there can be a trade-off between low regeneration energy and high cyclic working capacities; also when the \(CO_2\) concentration in the inlet stream is increased, the energy indexes tend to decrease (compensated by higher cyclic capacities). In the case of 30 wt% MEA + water, calculations show a cyclic working capacity in the range of 60–75 kg \(CO_2\) m\(^{-3}\) sln, and a corresponding...
regeneration energy of 3.2–2.8 MJ kg CO$_2$\(^{-1}\), in good agreement with literature estimates for a MEA-based CO$_2$ capture plant with a conventional process configuration (Boot-Handford et al., 2014). It can be seen in Figure 4 that a number of amines can achieve higher cyclic capacities or lower energy indexes than MEA. In addition, results show that for the same total amine mass concentration, non-aqueous amine solvents possess a 5–10% reduction in cyclic working capacity, with a 10–30% decrease in the energy index compared to their aqueous counterparts, mainly due to lower heat of vaporization and sensible heat (Alkhatib et al., 2019). The lowest regeneration energy was estimated for MEA + EGME and for systems with AMP, due to the previously discussed in Figure 3. Moreover, the higher temperature of regeneration of these solvents compared to their aqueous counterparts is not reflective of the expected lower temperature of regeneration, partly due to the shortcut method being developed for typical stripper configurations with water as cosolvent, and partly due to the comparison. A more accurate estimation of the temperature of regeneration would entail examining different process configurations for these low volatility cosolvents.

As already mentioned, the shortcut model represents ideal cases indicating an upper bound of performance, where heat and mass transfer are instantaneous and no temperature or concentration gradients exist (Ajenifuja et al., 2020). Moreover, accurate prediction of the energy index is dependent on reliable estimates of the specific heat capacity of the studied solvents. Since specific heat capacities were taken constant, predictions of energy index should be interpreted with caution. Nevertheless, the simple analysis implemented in this work allows a systematic comparison for initial solvent screening, as all of them have been studied under the same conditions. Nonetheless, more rigorous process modeling for these solvents, once screened for the most promising ones, are required for a holistic understanding of their performance, which are outside the scope of the current work. Other thermophysical properties, such as viscosity and surface tension, and other relevant properties to process design such degradation and evaporation should be included. For an excellent example on the implications of including transport properties in the cost analysis of absorbent materials, the reader is referred to a work by Mota-Martinez et al. (2017).

Furthermore, the presence of impurity gases such as SO\(_x\) and NO\(_x\) can greatly complicate the CO$_2$ separation process, producing unwanted reaction products, and/or degrading the solvents used for the separation (Mello and Eiü, 2002). For instance, in the benchmark absorption with MEA, the column cannot be fully regenerated because of irreversible degradation in reaction with oxygenated compounds that can produce foaming, fouling, and formation of unregenerable heat-stable salts in the amine, among others (Supap et al., 2006). SO$_2$ control is particularly important because its concentration (500–3,000 ppm) is higher than any other acid gas components in the flue gas (Rao and Rubin, 2002). Hence, a previous desulfurization process is often included in industrial recovery processes. Nevertheless, although deep desulfurization with >99% SO$_2$ removal is achievable, such an operation has economic implications (Lee et al., 2009). Moreover, regarding the modeling of impurities, the interactions of CO$_2$/H$_2$O/SO$_x$/NO$_x$ in typical flue gas with absorbents are quite complex. As pointed out by Lee et al. (2009), the information on the reaction rates of heat-stable salt formation between MEA and acid gases can rarely be found in the literature. Nevertheless, as an example of the capabilities of using a molecular-based model, CO$_2$ partial pressures in acid media were previously predicted in our group by using DEA, MEA, and MDEA via modeling with soft-SAFT (Alkhatib et al., 2019). Index of agreement values of 0.844, 0.766, and 0.705 were obtained, respectively, with differences in the performance of the model attributed to the availability of amines concentrations range used when optimizing the cross-association energy parameters for each acid gas–amine pair (Alkhatib et al.,

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**FIGURE 3** Comparison of calculated partial pressure of CO$_2$ as function of CO$_2$ absorption capacity (solubility) in aqueous and non-aqueous amine-based solutions (30 wt%) at different temperatures: (A) 313 K and (B) 393 K.
Novel Porous Solid Adsorbents for CO\textsubscript{2} Capture

In this section, a computational study of several adsorbent materials for CO\textsubscript{2} capture and separation is presented. Simulated adsorption isotherms were previously validated with experimental data in order to determine the accuracy of the models, as well as the quality of the multicomponent mixture predictions. Very good agreement was obtained for all pure component isotherms (AAD, \(\sim3\text{-}8\%\) for CO\textsubscript{2}, and \(<4\%\) for N\textsubscript{2}; Builes et al., 2011; Prats et al., 2017b; Bahamon et al., 2020), although some deviations above 11\% were obtained at higher pressures for MOFs with uncoordinated metal centers (e.g., M-MOF-74), due to limitations in the force fields available in the literature (Bahamon and Vega, 2016). A comparison of the adsorption isotherms for pure CO\textsubscript{2} at a temperature of 313 K is provided in Figure 5. As shown, the M-MOF-74 family displays a high affinity for carbon dioxide at low pressures (especially the structures with Mg and Ni as metal centers), followed by zeolites like NaX and NaY. Operation in the low-pressure range is beneficial for zeolites, which more easily attract the quadrupole present in CO\textsubscript{2}. However, when the operating pressures exceed atmospheric conditions, they become saturated due to their smaller pore volumes and because of the extra framework cations; higher uptake capacities can be then obtained with MOFs or ACs, instead. In addition, IRMOF-1 and MOF-177 display roughly linear trends of adsorption vs. pressure, indicating that the frameworks are not very selective for CO\textsubscript{2} adsorption and saturation capacity can be reached at higher pressure conditions.

Phenomena as coadsorption and/or site competition should be taken into account when evaluating process streams, since they can affect the performance of the materials for capturing CO\textsubscript{2} from an actual flue gas mixture (Bahamon et al., 2018); however, the amount of experimental data at these conditions is very scarce. Prediction of the mixture behavior can be obtained by the parametrization of pure isotherms, although the applicability can fail when preferential adsorption location at the intersections of the framework is present, as it is the case for highly selective materials for CO\textsubscript{2}-over-N\textsubscript{2} such as of NaX, UTSA-16, and Mg-MOF-74. Conversely, GCMC simulations allow direct calculation of the interactions between surface and molecules, explicitly including the adsorption competition between substances (Bahamon et al., 2020). This is one of the cases in which molecular simulations can be used as a robust, complementary, tool to the experiments.

In order to make a realistic comparison for gas streams at process conditions, results for selected materials are presented in Figure 6 for the same CO\textsubscript{2} concentrations as for absorption in amine-based solvents. It was observed that CuBTC and MOF-177 show poor performance due to competition with nitrogen, which reduces the adsorbent capacity (and therefore, the purity that can
be achieved in a swing adsorption cycle). In contrast, Mg-MOF-74 and zeolite NaX exhibit exceptional CO$_2$ storage capacity and selectivity (Bahamon et al., 2018). Moreover, in addition to the adsorption isotherms, GCMC also provides useful molecular insights into physical phenomena occurring at such operating conditions; hence, we present in Figure 6 some snapshots of CO$_2$/N$_2$ binary mixture for the different adsorbent families. It can be seen, for instance, that the honeycomb-like structure of Mg-MOF-74, with large one-dimensional pores and open-metal centers, is highly attractive for carbon dioxide molecules (especially with adsorption close to the metal centers).

Molecular simulations were further used to evaluate the behavior at operating conditions for temperature swing adsorption (TSA) processes, by assuming one can take advantage of low-grade thermal energy from the same industrial plant at which the CO$_2$ capture unit is going to be located. A simplified
model was used for the screening (Ajenifuja et al., 2020), as in the previous section with absorption. The cyclic working capacity was calculated as the actual amount $\Delta n$, recovered between adsorption and desorption conditions. Adsorption was obtained at 313 K and CO$_2$ compositions of 25, 15, and 5%, while desorption was obtained as pure CO$_2$ at temperatures up to 403 K:

$$\Delta n = n_{\text{ads}}(T, P_{\text{CO}_2}) - n_{\text{des}}(T, P) \quad (5)$$

In addition, the energy index (i.e., energy per amount recovered) was calculated considering the energy required to desorb the molecules from the solid material (also called isosteric heat $q_{\text{ST}}$) and the energy to heat those molecules up to the desorption temperature (i.e., sensible heat), as follows:

$$E_{\text{index}} = \frac{Q_{\text{sens}} + Q_{\text{ads}}}{\Delta n_{\text{CO}_2}} = \frac{(C_{\text{P}_{\text{CO}_2}} + C_{\text{P}_{\text{N}_2}})(T_{\text{ads}} - T_{\text{des}}) + \Delta n_{\text{CO}_2} q_{\text{ST}, \text{CO}_2} + \Delta n_{\text{N}_2} q_{\text{ST}, \text{N}_2}}{\Delta n_{\text{CO}_2}} \quad (6)$$

where $C_P$ are the heat capacities of the gases taken as an average constant value between adsorption and desorption temperatures.

A comparison of solid adsorbents in terms of cyclic working capacity and energy index is presented in Figure 7, analogous to the absorption case. Since in an adsorption process, both CO$_2$ and N$_2$ can be captured on the adsorbent material, a high selectivity for carbon dioxide is prerequisite in order to achieve high purities at the outlet, achieving similar specifications as for absorption. In this case, a minimum value of 90% CO$_2$ on the adsorbent was set as limit to compare the materials for industrial processes. In the literature, a few examples can be found utilizing theoretical methods to screen cost-effective materials for CO$_2$ capture (Krishna and van Baten, 2012; Lin et al., 2012; Wilmer et al., 2012; Maring and Webley, 2013; Huck et al., 2014; Dunstan et al., 2016), with typical adsorbent evaluation criteria such as structure–property relationships (e.g., surface area and pore volume), working CO$_2$ capacity, selectivity, or sorbent selection parameter. Nevertheless, as the case of absorbents, the proper metric for comparing different adsorbents should be a volumetric basis, taking the bed volume and space requirements of the facilities as constraints. Furthermore, it is worth mentioning that because the gas stream is typically discharged at near atmospheric conditions, compressing or extracting a vacuum to such huge amounts of the dry gas stream is predicted to be more costly than heating the solid material (Hedin et al., 2013). A fair comparison of energy requirements between the two technologies studied here (adsorption and adsorption) is presented using the same thermal energy concept. In this case, the crystal densities of the materials were used, with the void fraction of the bed held constant and equal to 0.4.

As shown in Figures 7, 4 of the 16 adsorbent materials examined (including zeolite NaX) are able to achieve low energy indexes with high cyclic working capacities (lower right quadrant) for 25% CO$_2$, as well as to satisfy the purity specifications. If the concentration is reduced to 15% CO$_2$, only three of these adsorbents are able to meet the performance specifications, and none of the materials is capable of achieving concentrations over 90% CO$_2$ on the adsorbent starting with a stream with 5% CO$_2$. M-MOF-74 and zeolites NaX and NaY offer the best results, demonstrating that these materials can be appropriate for this TSA technology. Adsorption-based processes with values as low as 1.45 MJ kg CO$_2^{-1}$ have been
obtained, without including the waste heat required to build steam [estimated in 1.26 MJ kg CO$_2$$^{-1}$ (Joss et al., 2017)]. In similar screening exercises for post-combustion carbon capture by VSA processes, UTSA-16 and Mg-MOF-74 were identified as the top-ranked materials by energy index (Khurana and Farooq, 2016), showing the tendency to adapt to changes in T,P of the latter. It should be mentioned that materials with less affinity for carbon dioxide such as ACs, MCM-41, etc. present low cyclic working capacities for the condition presented; however, higher adsorption pressures will result in a much higher amount of CO$_2$ recovered with these materials.

As seen, useful insights about the sensitivity of adsorbent performance to changes in operating conditions can be extracted. In some cases, higher working capacities than the ones obtained for the minimum of the energy index are achievable with a slight increase of $\sim$10% in the energy requirements (see dashed spheres in Figure 7). However, this comes as an expense of higher temperatures needed for regeneration. Nevertheless, this characteristic of some adsorbents must be addressed. Since less cycles are required to capture the same amount of CO$_2$, the whole capture process is faster and cheaper (Vujic and Lyubartsev, 2017). Furthermore, the election of the desorption temperature must be counterbalanced with the energy demand of the process (Plaza et al., 2017) and with the cost of heating the bed depending on the quality of heating required. If higher pressure steam were used for heating, which is a more valuable utility, higher desorption temperatures, and higher working capacities might be achieved (Ajenifuja et al., 2020). In addition, as the adsorbent is more selective for CO$_2$ than N$_2$, CO$_2$ uptake is more sensitive to increases in the desorption temperature, thus leading to an increase in CO$_2$ purity.

From an engineering point of view, swing adsorption cycles can be fine-tuned to fulfill a variety of demanding requirements. Increasing the complexity of the cycle by adding additional recycle, purge, secondary heating steps, heat recovery, or low adsorption temperatures by using chilled water or refrigerants could improve the performance of the adsorbent (Grande and Rodrigues, 2008; Radosz et al., 2008; Plaza et al., 2010; Ntiamoah et al., 2016; Sinha et al., 2017; Zhao et al., 2019), but such integration of the process is out of the scope of this work. In addition, with the assumption of equilibrium adsorption in the model, cycle productivity cannot be directly estimated. Consideration of heat-transfer rates and other transport properties, although important for the detailed design, is not critical for the initial screening of candidate adsorbents under process conditions (Ajenifuja et al., 2020), although detailed process optimization will definitely provide the most accurate ranking (Balashankar et al., 2019).

Finally, it is worth mentioning that some adsorbent structures achieve similar values to the ones obtained in the previous section for alternative amine-based systems (2–2.5 MJ kg CO$_2$$^{-1}$). Nevertheless, TSA processes present the disadvantage of the time delay imposed by the heating and cooling steps and the purity reached under very low CO$_2$ concentrations in the inlet stream. Further improvements in the cycle time process optimization are required, intrinsically associated with the feed flowrate, the height of the adsorber, adsorbent characteristics, and the cycle design, among others. Its future applicability will rely upon the stability of the novel material and the price.

In addition, capturing CO$_2$ from typical flue gases encompasses an additional difficult related to the presence of water, NOx, SOx, and other impurities, which are detrimental to most CO$_2$-selective adsorbents. It is well known that water adsorption excludes CO$_2$ from the adsorption sites and hence could potentially annul all separation capacity of the adsorbent (Zhang et al., 2009). For instance, the equilibrium adsorption capacity of CO$_2$ on NaX decreased by 99% when using untreated flue gas (i.e., in the presence of H$_2$O; Li et al., 2008). Therefore, it is necessary to use prepurification units or prelayers of adsorbent with excellent water working capacity for effective CO$_2$/H$_2$O swing adsorption processes (Cavenati et al., 2006). Activated alumina and silica gel are common adsorbents used in industries for removing water vapor (Kumar et al., 2003; Serbezov, 2003). As aforementioned for the adsorption process, such pretreatment stages could considerably increase the overall cost. With respect to molecular simulation studies focused on this endeavor, Huang et al. (2012) found that the interaction between water molecules and the framework plays a crucial role when evaluating the effect of a trace amount of water on CO$_2$ capture. Liu et al. (2014) showed that the presence of H$_2$O reduces the CO$_2$ adsorption ability but increases the CO$_2$/N$_2$ separation factor. In a previous work from our group, the impact of the impurities on the performance of the materials was evaluated at the molecular level and process conditions for selected MOFs and comparing with the benchmark NaX material (Bahamon et al., 2018). Results showed that the zeolite is a more hydrophilic adsorbent than CuBTC and Mg-MOF-74, becoming useless with $<0.1%$ of water content in the mixture, and just a 0.01% of moisture is able to significantly reduce the CO$_2$ adsorption capacity. Moreover, the energy requirements grows exponentially with the inclusion of H$_2$O and SO$_2$, although it is remarkable that there are “sweet spots” in MOF-74 where the use of small traces of impurities allow an enhancement in the KPIs due to the competition with nitrogen for the remaining adsorption sites (Bahamon et al., 2018). Nevertheless, many fundamental works need to be done to gain a better understanding on the mechanism of these interactions and their effect on the process performance (Zhang et al., 2009) to be further transferred to the proposed methodology used in this contribution. Production of ad hoc experimental data is required for validating purposes.

CONCLUSION

In this work, we have shown, through some specific cases, how molecular modeling techniques (specifically soft-SAFT and GCMC) can be used to obtain thermophysical properties of alternative selected solvents and adsorbents for CO$_2$ capture, allowing a systematic comparison of the effect of the molecular structure on the performance of both technologies. The cyclic working capacity and energy index were used as the two KPIs to evaluate the performance of the selected systems under the same operating conditions. In the case of solvents, aqueous MEA was used as the benchmark, systematically searching for
alternatives requiring lower energy index with higher cyclic working capacities. We have obtained that, for same total amine mass concentration, non-aqueous amine solvents provide a 5–10% reduction in cyclic working capacity, with a 10–30% decrease in the energy index compared to their aqueous counterparts. Sixteen adsorbents, belonging to different types of materials, were also compared under the same conditions, obtaining that M-MOF-74 and zeolites NaX and NaY provide the best results for adsorption in TSA processes. A comparison between the different cases studied for solvents and adsorbents show that some adsorbent materials need similar values of energy requirement to the ones obtained for alternative amine-based systems (2–2.5 MJ kg CO$_2$ $^{-1}$), although the disadvantage of the TSA process versus absorption should be taken into account, as already explained.

Overall, this work shows that a simplified analysis can be used to guide the development and selection of the most promising systems for CO$_2$ capture, based on key process parameters. A wide range of operating conditions can be studied, and useful insights into the impact of operating conditions on the performance can be gained with this approach. Considering the high degree of transferability of the model parameters, the applied approach stands as a robust and valuable/effective tool for the screening and discovery of new possible systems for carbon capture. Further work will include extensions to hypothetical systems (not evaluated experimentally yet) and implementation in engineering simulators for process simulation and design. It has been shown that CO$_2$ capture technologies, with lower energy requirements than the current ones, can be implemented in the industrial sector. However, in combination with the Paris Agreement climate ambition, establishing markets for premium lower-carbon materials can also accelerate the adoption of such CCUS processes.

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DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

AUTHOR CONTRIBUTIONS

DB is a research scientist who carried out some of the GCMC simulations work, data analyses and interpretation, and writing of the manuscript. NA is a research assistant who performed the GCMC calculations for MOFs at different concentrations. SB is an assistant professor who performed GCMC simulations of silica-based and functionalized silica-based materials. IA is a Ph.D. student who performed the soft-SAF calculations and interpretation. MS is an assistant professor who helped in the design and interpretation of the simulation work. LV is a professor and the academic supervisor of the work; together with DB, she defined the strategy of the work, followed up on the results and interpretation and revised the overall work. All authors have contributed in the drafting and the revision of the article.

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