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FULL PAPER

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Alkalinity Concentration Swing for Direct Air Capture of Carbon Dioxide

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The Alkalinity Concentration Swing (ACS) is a new process for direct air capture of carbon dioxide driven by concentrating an alkaline solution that has been exposed to the atmosphere and loaded with dissolved inorganic carbon. Upon concentration, the partial pressure of CO₂ increases, allowing for extraction and compression. Higher concentration factors result in proportionally higher outgassing pressure, and higher initial alkalinity concentrations at the same concentration factor outgas a higher concentration of CO₂. Two desalination technologies, reverse osmosis and capacitive deionization, are examined as possible ACS implementations, and two corresponding energy models are evaluated. The ACS is compared to incumbent technologies and estimates for water, land, and energy requirements for capturing one million tonnes of CO₂ per year are made. Estimates for the lower end of the energy range for both approaches compare favorably to other approaches, such as solid sorbent and calcining methods.

KEYWORDS

Carbon, carbon dioxide removal, direct air capture, reverse osmosis, capacitive deionization, sustainable chemistry

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1 Introduction

Removal of carbon dioxide from the atmosphere has been proposed as an essential method for responding to anthropogenic climate change.^[1,2] Policymakers and scientists agree that in order to minimize future harm to society — which will be most felt by the world's most vulnerable populations — priority should be devoted to efforts and technologies that reduce emissions from burning fossil fuels and other sources of greenhouse gases.^[3] But even after deep decarbonization efforts, some hard-to-avoid emissions will remain, either because they are unacceptable to avoid from a social-justice perspective (e.g. food security constraints) or extremely physically difficult to eliminate within the given timeframe, making some degree of carbon dioxide removal (CDR) necessary.^[4,5] A gigatonne-per-year scale of global CDR will likely be required by the end of the century. However, aiming for larger scales, up to 20 billion tonnes of CO₂ per year (GtCO₂/year) as some reports suggest,^[6] has significant associated moral hazards and ethical considerations, as the promise of future CDR could deter decarbonization in the short term.^[7,8]

1.1 Carbon dioxide removal

Carbon dioxide removal spans a wide range of approaches, each with different associated materials, energy, land, resource, and societal consideration. Biological CDR methods — including reforestation^[9,10] and soil carbon management^[11] — are projected to be able to achieve gigatonne-scale per year removal,^[6] though they tend to store carbon in impermanent reservoirs, meaning they are more susceptible to reversals. While these approaches confer co-benefits, such as increasing biodiversity and improving local water and soil quality, they also require significant land area to reach gigatonne scale and may compete with other land-use demands, such as agriculture or conservation objectives.^[2]

Other approaches that store carbon in a more durable form include bioenergy with carbon capture and storage,^[12] carbon mineralization processes that remove CO₂ directly out of the air,^[13] or the addition of alkalinity to oceans, which increases dissolved carbon and ultimately drives the production of carbonate sediments.^[14,15,16,17] Mineralization processes could result in the long-term storage of concentrated CO₂ streams in subsurface formations, products such as concrete, as well

as mine tailings and alkaline industrial wastes.^[6,18]

1.2 Industrial direct air capture approaches

An alternate strategy for carbon dioxide removal involves direct air capture (DAC), industrial technologies for separating atmospheric CO₂ directly from the air through chemical or physical processes,^[19] coupled with sequestration (e.g., storage in a geological reservoir or through mineralization).

One class of approaches is based on solid sorbent technologies that typically use amine-based materials to reversibly bind CO₂. This process can be cycled many times to capture CO₂ out of ambient air and release a concentrated stream through a thermal^[19] or moisture swing.^[20] Alternatively, recent work demonstrated a faradaic electro-swing adsorption system, which uses voltage to regenerate CO₂.^[21]

Another class of approaches relies on a basic aqueous solution to absorb CO₂ from ambient air. One commercialized process is based on an aqueous potassium hydroxide contactor that absorbs CO₂ directly from air and converts the CO₂ to calcium carbonate; releasing the CO₂ requires heating of calcium carbonate to approximately 900°C.^[22] A different approach makes use of an electrochemical swing, which changes the pH of the solution and allows for the release of CO₂ without going through the steps of precipitation and heating.^[23]

In this paper, we describe a new DAC approach that is based on taking a dilute alkaline aqueous solution that has equilibrated with air, and concentrating it. Concentrating the alkalinity and the dissolved carbon increases the partial pressure of CO₂ in the solution and allows for CO₂ outgassing and extraction. We describe the chemical cycle underlying this approach, evaluate the thermodynamics of the process, and examine two commercially available technologies that are traditionally used for desalination, reverse osmosis and capacitive deionization, to drive the cycle. We then compare the potential advantages of this approach relative to other existing DAC methods and analyze its scale-up feasibility.

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2 The Alkalinity Concentration Swing for direct air capture

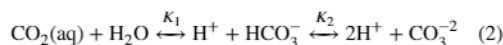
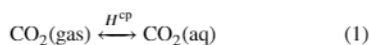
Our new approach for direct air capture is based on the recognition that an alkaline aqueous solution containing an air-equilibrated concentration of dissolved inorganic carbon (DIC) — the sum of carbonate ion (CO_3^{2-}), bicarbonate ion (HCO_3^-), and dissolved aqueous carbon dioxide ($\text{CO}_2(\text{aq})$) concentrations in solution — will release CO_2 to the air when that solution is concentrated. After outgassing, if the same alkaline aqueous solution is diluted, CO_2 absorbs from the air and the DIC concentration increases. We use this phenomenon as the core component of the “Alkalinity Concentration Swing” (ACS) cycle for capturing atmospheric CO_2 . In this section we describe the basis for this approach and present a set of idealized steps for realizing the ACS.

2.1 The equilibrium aqueous carbonate system for varied alkalinity

The concentration of DIC in equilibrium with atmospheric CO_2 ($p_{\text{CO}_2} \approx 0.4$ mbar; equivalent to roughly 400 ppm) depends on the alkalinity — the molar charge difference between the sum of the conservative cations and that of the conservative anions, i.e. ions whose concentrations do not vary with pH. For simplicity, this work assumes alkalinity to be the moles of K^+ ions per liter. As the alkalinity of the solution increases and equilibrium with air is maintained, the pH increases and the amount of DIC increases, but at a decreasing rate. This is due to the transition of the dominant species of DIC from bicarbonate to carbonate as pH increases above ~ 9.5 . Indeed, in very dilute solutions (alkalinity $< 1 \times 10^{-2}$ M), the slope of the DIC-alkalinity line is close to unity (Figure 1A inset), where each unit of alkalinity, or conservative cations, is balanced by monovalent bicarbonate ions. At alkalinity > 0.1 M, the slope is closer to 0.5 (Figure 1A main plot) and the charge balance required by this increase in alkalinity is accommodated primarily by the carbonate ion, which is divalent.

The DIC to alkalinity relationship follows from carbonate and aqueous chemistry equilibrium relations, as well as the charge-neutrality condition requiring that the excess charge of conservative cations over conservative anions equal the ex-

cess charge of non-conservative anions over non-conservative cations (see SI Section A.2). The relative ratios of carbon species in equilibrium is set by the following chemical reactions:



The following system of equations determines the relationship between CO_2 partial pressure, alkalinity, and DIC:

$$[\text{CO}_2]_{\text{aq}} = H^{\text{cp}} p_{\text{CO}_2} \quad (3)$$

$$K_1 = [\text{HCO}_3^-] [\text{H}^+] / [\text{CO}_2]_{\text{aq}} \quad (4)$$

$$K_2 = [\text{CO}_3^{2-}] [\text{H}^+] / [\text{HCO}_3^-] \quad (5)$$

$$K_w = [\text{H}^+] [\text{OH}^-] \quad (6)$$

$$A = [\text{HCO}_3^-] + 2 [\text{CO}_3^{2-}] + [\text{OH}^-] - [\text{H}^+] \quad (7)$$

Here, A is the alkalinity concentration in units of moles per liter. DIC concentration is defined as $C_{\text{DIC}} \equiv [\text{CO}_2]_{\text{aq}} + [\text{HCO}_3^-] + [\text{CO}_3^{2-}]$. At a fixed temperature, K_1 and K_2 vary slightly with the ionic strength^[24] and hydrostatic pressure^[25,26] of the solution. Across the studied conditions, these effects modestly enhance the efficiency of the ACS, but are neglected in this study for simplicity; equilibrium constants are fixed for pure water conditions at 20°C and 101.325 kPa: $K_1 = 9.6 \times 10^{-7}$ M, $K_2 = 3.4 \times 10^{-10}$ M.^[24] Henry’s constant, H^{cp} , is 0.034 M/bar and the water disassociation constant, K_w , is 1×10^{-14} M².

The decrease of the ratio of DIC to alkalinity with increasing alkalinity is the principle underlying the ACS. Consider a dilute solution of any strong base initially in equilibrium with air. If that solution is isolated from air and concentrated, for example through the removal of water, both the DIC and

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the alkalinity increase linearly in proportion to their relative concentrations in the solution. The DIC of a solution with the same corresponding alkalinity, but maintained in equilibrium with air, increases more slowly than that of the concentrated solution. Thus, the concentrated solution is supersaturated and spontaneously outgasses, which allows for the extraction of CO_2 .

This process is depicted in Figure 1, in which a solution with an initial alkalinity of 10 mM is concentrated by a factor of 100. When the solution reaches 1 M, p_{CO_2} becomes 40 mbar, which is a hundredfold increase over that in air. When the CO_2 outgasses, restoring equilibrium with air, 0.3 moles of DIC per liter of concentrated solution have been captured (Figure 1B). Relative to the initial feed solution, 3 mM of CO_2 or 35% of the DIC has been outgassed.

2.2 The Alkalinity Concentration Swing cycle

The following is an idealized description of the ACS based entirely on equilibrium aqueous carbonate assumptions described in Section 2.1 above. Specific methods for implementing the ACS and associated energetics are discussed in Sections 3 and 4.

2.2.1 Step 1→2: Concentrating alkalinity

A solution with initial alkalinity A_i is at equilibrium with the atmosphere at a given partial pressure of CO_2 , $p_i \approx 0.4$ mbar (State 1). The fraction of carbon species in solution and DIC concentration is set by A_i and p_i based on the aqueous carbon chemistry relations described in Section 2.1 above: $C_{\text{DIC},1} = C_{\text{DIC}}(A_i, p_i)$ (see SI Section A.2 for full derivation).

The system is then closed off from exchange with the atmosphere and the solution is concentrated such that the new effective alkalinity and DIC concentrations are increased by a concentration factor, χ (Figure 1; Concentrating step). Such a concentrating step does not change the absolute number of alkaline carrier ions or DIC molecules in solution, but increases the concentration of both by confining the solutes in a smaller volume. This is equivalent to removing solvent water molecules from solution. The alkalinity and DIC concentrations in the concentrated state are given by: $A_f = A_i \chi$ and $C_{\text{DIC},2} = C_{\text{DIC},1} \chi$, respectively (State 2).

2.2.2 Step 2→3: CO_2 outgassing

Once the system is in the concentrated state at the higher concentration of alkalinity, A_f , the aqueous CO_2 activity increases such that its equilibrium partial pressure rises to p_2 (State 2). In engineered systems, CO_2 will generally be collected from the concentrated solution by exposing it to a fixed outgassing pressure, p_f that is lower than p_2 (which we also refer to as $p_{f,\text{max}}$).

Exposing the system to p_f in the concentrated state drives the following disproportionation reaction:



Outgassing occurs as shown in Figure 1A-B. The concentration of DIC outgassed as CO_2 with respect to the feed solution as a result of the ACS is given by the following relationship (see SI Section B.1 for full derivation):

$$C_{\text{out}} = C_{\text{DIC}}(A_i, p_i) - \frac{A_i}{A_f} C_{\text{DIC}}(A_f, p_f) \quad (9)$$

The fraction of DIC species outgassed is given by:

$$f_{\text{out}} = C_{\text{out}}/C_{\text{DIC}}(A_i, p_i) \quad (10)$$

The upper limit of f_{out} is 0.5; it occurs only if the initial DIC is entirely made up of bicarbonate ions, and so the alkalinity to DIC ratio is exactly 1:1. If such a system is concentrated to a point where the DIC equilibrium shifts essentially entirely to carbonate at high alkalinity, in the 2:1 alkalinity to DIC regime, then half of the bicarbonate ions are converted to carbonate ions, and the other half become carbon dioxide molecules, which may be collected. In practice, the alkalinity to DIC ratio will fall between 1 and 2.

The maximum pressure at which outgassed CO_2 can be removed from the system is also of interest. Over the range of initial alkalinity between 10^{-4} and 10 M, the outgassing pressure limit is independent of initial alkalinity and is a direct relationship between the concentrating factor, χ , and the initial pressure, p_i , given by (see SI Sections A.3 and B.2 for full derivation):

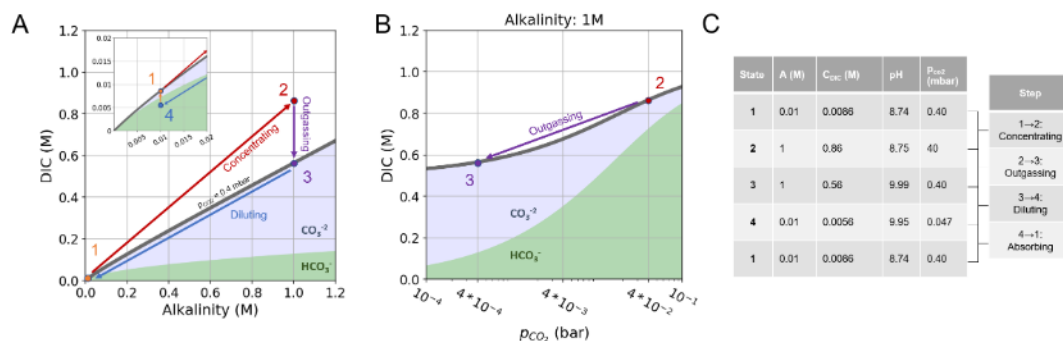


FIGURE 1 An Alkalinity Concentration Swing Cycle. (A) The gray line plots the dissolved inorganic carbon (DIC) concentration in equilibrium with atmospheric CO_2 ($p_i = 0.4$ mbar) as a function of alkalinity. The red arrow indicates the concentration step of the ACS and plots the trajectory when a solution, initially equilibrated at 0.01 M alkalinity (orange point), is concentrated by a factor of 100 to an alkalinity of 1 M. In the concentrated state (red point), the solution has excess DIC over that in equilibrium with air. The purple arrow indicates the amount of CO_2 that outgasses as the system reaches a new equilibrium at high alkalinity and $p_f = 0.4$ mbar. The remaining blue and orange arrows indicate the dilution and CO_2 absorption steps, which return the system to the initial state. Vertical extent of green and purple lavender regions indicate concentrations of bicarbonate and carbonate, respectively. Inset: The black curve plots the same DIC versus alkalinity relationship as the main plot, but from 0 to 0.02 M alkalinity, showing the transition between a roughly 1:1 alkalinity to DIC relationship at low alkalinity to a 2:1 scaling at higher alkalinity. (B) A plot of DIC as a function of p_{CO_2} at a fixed alkalinity of 1 M. The red and purple dots, as well as the purple arrow, correspond to panel A. (C) The alkalinity, C_{DIC} , pH, and p_{CO_2} values are listed corresponding to each state in (A).

$$P_{f,\max} \approx P_i \chi \quad (11)$$

Step 2→3 is concluded once the system has reached its new equilibrium point at A_f and p_f , setting a DIC concentration of $C_{\text{DIC},3}$ (State 3).

2.2.3 Step 3→4: Diluting alkalinity

The next step of the ACS involves returning the concentrated alkalinity, A_f , to its initial value. This can be done by recombining the concentrated solution with the removed water from Step 1→2. Alkalinity is diluted by a factor of $1/\chi$ to A_i and DIC is diluted by the same factor giving $C_{\text{DIC},4} = C_{\text{DIC},3}/\chi$ (State 4).

2.2.4 Step 4→1: Absorption of atmospheric CO_2

The final step, which returns the system to State 1, exposes the solution to the atmosphere. Absorption occurs because the dilution step has created a condition with less DIC relative to the concentration in equilibrium with the atmosphere. CO_2

is consumed via the comproportionation reaction: $\text{CO}_2(\text{aq}) + \text{CO}_3^{2-} + \text{H}_2\text{O} \rightarrow 2\text{HCO}_3^-$ (or, the reverse of disproportionation). Step 4→1 is concluded once the system returns to its equilibrium point at A_i and p_i . Steps 3→4 and 4→1 could occur simultaneously, in principle, as could Steps 1→2 and 2→3.

2.3 CO_2 Outgassed from the ACS

For a given temperature, the exact choices of parameters for the initial and final alkalinity values, as well as initial and final CO_2 partial pressures, uniquely determine the outputs of the ACS. Whereas the initial pressure is set by the concentration of atmospheric CO_2 , the outgassing pressure is a design parameter that should be set based on considerations relating to energy, rate, and water requirements. In this study, we fix the outgassing pressure at $p_f = 0.4$ mbar. The energetic considerations of the outgassing pressure are discussed briefly in Section 4.1, and will be the subject of future studies.

Figure 2 plots the result of the ACS for a fixed atmospheric and outgassing partial pressures of CO_2 over a range of initial alkalinity values. The concentration factor specifies the

concentration (C_{out} ; Equation 9) and fraction (f_{out} ; Equation 10) of DIC outgassed as CO_2 . Outgassing purity thresholds are calculated based on partial pressures of other atmospheric gases (N_2 , O_2 , Ar; see SI Section B.4); higher concentration factors yield higher CO_2 purity. In general, higher initial alkalinity values for the same concentration factor yield larger total outgassing values. The fraction of DIC outgassed exhibits a more complicated relationship with concentration factor. The lower the initial alkalinity, the higher the outgassed fraction can be (with an absolute limit at 0.5); the limiting regime is set when all DIC is in bicarbonate form and entirely disproportionates. As given by Equation 11, the maximum outgassing pressure is set only by the concentration factor, invariant of the initial alkalinity. Increasing the concentration factor therefore increases the difference between the outgassing pressure (p_f) and the partial pressure of the solution in the concentrated state (p_2), which corresponds to higher absorption rates.^[27] Table 1 lists output values for different representative ACS input parameters.

3 Implementing the ACS

The primary energy-consuming driving mechanism behind the ACS can be separated into two components: 1) a process to concentrate solutes in water, and 2) applying pressure for outgassing of CO_2 from solution (Figure 3A). The remaining components, diluting alkalinity and absorbing CO_2 , do not consume energy but are critical for evaluating water and contacting area requirements, as described in Section 5.

In principle, any desalination method, which produces purified water, can also be used to concentrate a stream of solute-filled solution. Desalination methods, for this reason, are candidate drivers for the ACS; they can be based on the following mechanisms: reverse osmosis (RO),^[28,29] capacitive deionization (CDI),^[30] electrodialysis,^[31] evaporation and distillation,^[32,33] precipitation,^[34] and solvent solubility.^[35] In this study, RO and CDI are considered for implementing the ACS. Both methods are deployed on an industrial scale, and evaluating the two serves as a comparison between pressure driven and electric field driven approaches.^[36] Energetics of the ACS process using these two processes as examples are discussed in Section 4.

RO is a membrane-based separation process in which pres-

sure is applied against a solvent-filled solution, overcoming the osmotic pressure of the solution, to create a concentrated and a dilute stream (Figure 3B). RO methods can be applied to brackish (low salinity) waters and wastewater processing with more dilute solutions, but are most commonly applied to seawater desalination. This application of RO is in broad commercial use, producing roughly 100 million cubic meters of purified water per day in 2018.^[37] Seawater desalination plants are typically designed to produce a stream of freshwater from an input feed of about 0.6 M of NaCl equivalent salt, yielding a concentrate output of roughly double the original salinity. In general, the RO process can be adapted to a broader range of initial salinities and higher overall concentration factors that may be desirable to achieve more optimal ACS output.

Existing technological developments and future prospects make RO an appealing candidate for implementing the ACS. For example, the development of energy recovery devices (ERDs) was crucial in reducing the power consumption of desalination to its current level. ERDs use the remaining energy stored in the pressure of the concentrate, which otherwise would be wasted, to apply part of the necessary pressure to the feed.^[28] One significant difference between an ACS process and desalination is that in the ACS, after CO_2 has been extracted, the concentrated and diluted solution are recombined. It is therefore possible to recover some of the energy held in the salinity gradient between the concentrated and dilute streams through forward osmosis,^[38] as well as reverse electrodialysis^[39,40] (see SI Sections D.3.3 and D.3.4 for analytic treatment). Chemical treatment of RO membranes can modify selectivity to gases, such as N_2 and CO_2 ,^[41] which will affect efficiency of the ACS-RO process.

An alternative approach to the concentration step for ACS is CDI, which is a method of concentrating and removing anions and cations from solution by applying a voltage across two electrodes and creating an electric double layer made up of electrolyte ions (Figure 3C). When voltage is applied (<1.2 V to avoid splitting water), anions “electrosorb” to the positive electrode and cations to the negative electrode. When the voltage is switched off or reversed, the concentration of ions in the electrode pores and in the fluid between the electrodes sets the output concentration of a higher-alkalinity solution, driving the concentration step of the ACS. The material properties of

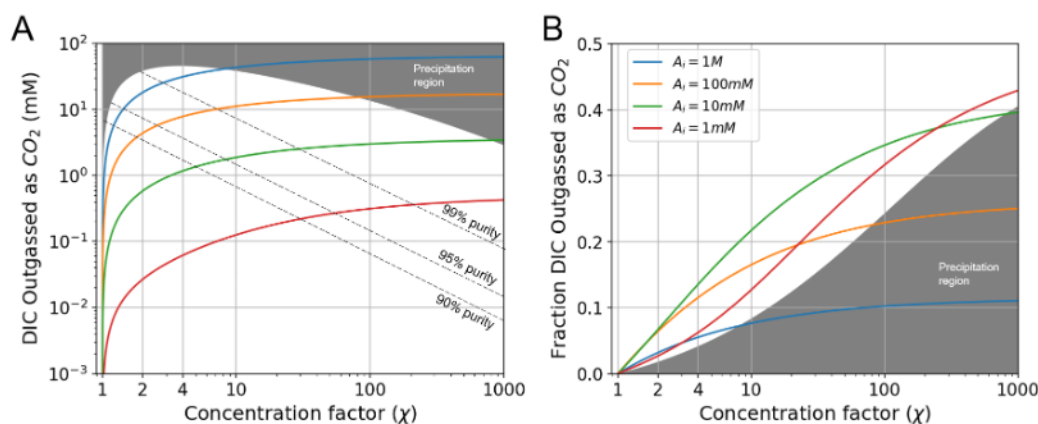


FIGURE 2 DIC outgassed based on the ACS. (A) Concentration of DIC in the feed solution outgassed as CO₂ as a function of concentration factor. Dashed lines represent purity thresholds with respect to non-CO₂ gases. (B) The fraction of DIC in the feed solution outgassed as CO₂ as a function of concentration factor. Each color curve represents a different initial alkalinity concentration (legend in panel B applies also to panel A). The boundary of the gray region represents the precipitation threshold for potassium carbonate at 20°C of approximately 8M.

the electrode (surface area, porosity, surface chemical groups, etc.) and the geometry constrain the overall capacity, rates, and energies of deionization.^[30] To increase efficiency, ion exchange membranes can be placed between the feed solution channel and the electrode material, in which case the process is called membrane CDI (MCDI).

Whatever approach is used to concentrate the alkaline solution, CO₂ can be extracted from the concentrate stream by exposing it to a vacuum or a carrier gas. This can be done through a variety of standard methods in chemical engineering including vacuum pumps, or by making use of water vapor or another condensable gas (e.g. helium). Using a liquid-gas exchange membrane for CO₂ extraction,^[42,43] which creates a gas-permeable barrier between gas and liquid phases, allows for lowering the outgassing pressure (<1 mbar) significantly below the vapor pressure of water while preventing flash evaporation.

Once CO₂ is extracted from solution, the concentrated and dilute streams are combined, thereby diluting alkalinity to its initial concentration. At this point, the solution has less DIC relative to alkalinity than it would have at atmospheric conditions. Exposing this solution to the atmosphere will initiate an equilibration process of CO₂ absorption, which is

critical to evaluate in order to assess ACS requirements for water processing, water on hand, and land use. Due to the relatively low hydroxide ion concentrations associated with ACS conditions, air-liquid contactors, which increase the surface area of solution and use fans to increase exposure to air, are likely to be ineffective as a result of slow absorption kinetics. Instead, we can use large contacting pools, potentially with mechanisms to enhance convective mixing. The kinetics of gaseous CO₂ absorbing into water and reacting with hydroxide ions to form bicarbonate ions is the rate limiting step in the absorption process.^[44] Overall absorption rate increases like the square root for higher hydroxide concentrations (i.e. higher pH) and linearly for higher air-liquid surface area. In Section 5.2, we use an approximated absorption rate to estimate the water on hand requirement for an ACS system given a certain facility water processing rate. The absorption time scale sets the duration that processed water needs to reside in the pool to reload DIC back into solution, and thus sets the total amount of water needed in the pool to operate the system in a continuous manner.

In the next section, we discuss the energy requirements of the ACS including the work associated with CO₂ extraction. We introduce two simple energy models to explore ACS en-

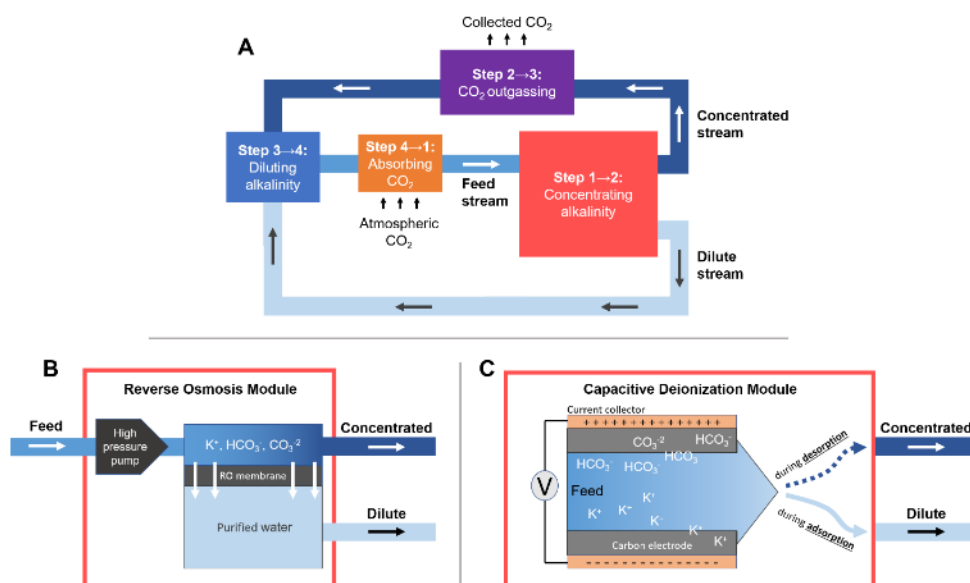


FIGURE 3 ACS system schematic. (A) The four steps of the ACS represented in a full cycle. (B) A schematic of a reverse osmosis module driven by a high-pressure pump. (C) A schematic of a capacitive deionization module driven by applied current and voltage. Ion exchange membranes are not represented in the diagram.

ergy trade-offs, referencing a range of energy values based on reported values from RO and CDI systems.

4 ACS thermodynamics and energy models

The process of concentrating ions in aqueous solution can be achieved, in general, by doing work to either confine ions to a smaller volume or to selectively remove water molecules from solution. Whereas the fundamental thermodynamic limit for the work required by a concentrating process is set by the entropic difference between the input and output streams, the particular mechanism for concentrating determines the additional associated dissipated energy.

This section describes the thermodynamic minimum work of the ACS, discusses the associated vacuum outgassing energy requirements, and explores two high-level frameworks for evaluating the energetics of concentrating ions in solution to achieve Step 1→2 of the ACS. Two simplified energy models are proposed, one based on energy associated with reverse osmosis and another based on energy of binding ions

in solution. Reverse osmosis and capacitive deionization are discussed as possible implementations of systems capable of concentrating ions to drive the ACS.

Irrespective of the particular concentrating mechanism, it is possible to set a thermodynamic limit on the ACS given an input and output partial pressure of CO₂. If the ACS cycle takes an input partial pressure of CO₂, p_1 , and outgasses at a limiting output pressure of $p_{f,max}$, the thermodynamic minimum work per mole CO₂ is given by: $w_{lim} = RT \ln(p_{f,max}/p_1)$, as long as it behaves as an ideal gas. Using carbonate chemistry assumptions (based on Equation 11), we rewrite the thermodynamic minimum work expression in terms of the concentration factor, χ , as:

$$w_{lim} = RT \ln(\chi) \quad (12)$$

If no vacuum is applied, a concentration factor of greater than 2500 — which is approximately equivalent to the ratio of non-CO₂ molecules to CO₂ molecules in the input stream of atmospheric air — is needed to outgas CO₂ at 1 bar.

4.1 Work needed for CO₂ extraction from an aqueous solution

The nature of the ACS is such that the CO₂ partial pressure limit in the concentrated state over the parameter range of interest (alkalinity between 0.1 mM to 10 M) is essentially proportional to the concentration factor (Equation 11). In order to extract CO₂ out of the concentrated solution, a CO₂ partial pressure that is less than $p_{f,max}$ must be chosen. If the concentration factor is lower than ~ 2500 , a vacuum could be applied to the solution to extract CO₂, or the headspace could be filled with a separable carrier gas (e.g. helium or water vapor). For simplicity we analyze the application of vacuum.

The thermodynamic minimum work needed to establish a vacuum at p_f to permit CO₂ outgassing isothermally is $w_{vac,min} = RT \ln(p_0/p_f)$, where $p_0 = 1$ bar; if $p_f = 0.4$ mbar, then the minimum work is 19.1 kJ/molCO₂ (abbreviated as kJ/mol whenever referring to moles of CO₂). However, real physical systems incur additional losses from dissipation. Industrial vacuum pumps have process efficiencies in the range of 65-85%.^[45]

There is a trade-off between the work needed to concentrate the feed solution and the work needed to establish a vacuum for extracting CO₂ from solution. In this study, the work needed to establish the vacuum is held constant because we have chosen to fix the outgassing pressure, p_f , at 0.4 mbar, and subsequently bring the outgassed CO₂ gas to 1 bar. Assuming an efficiency of $\eta = 70\%$ yields an additional work of $w_{vac} = w_{vac,min}/\eta \approx 30$ kJ/mol. The trade-off that emerges from varying the outgassing pressure is a subject for future study.

Finally, because we assume the thermodynamic limit of isothermal compression, there is no additional required work to compress water vapor. Assuming it begins at its equilibrium vapor pressure at 20°C of 40 mbar, it precipitates during isothermal compression with no additional work, and its partial pressure remains at 40 mbar as the CO₂ is compressed to a partial pressure of 1 bar.

4.2 Reverse osmosis-driven ACS

4.2.1 Energy model

We first pose a model for the work required to concentrate a solution through confining ions to a smaller volume. For an

aqueous solution, when entropic effects are dominant, the relevant macroscopic state variables that determine free energy as the solution is concentrated and diluted are the osmotic pressure and concentration. The change in these state variables through the ACS sets the work necessary to concentrate the feed solution so that CO₂ can be extracted.

In this "RO model," we assume an idealized ion concentrating RO system in which water is driven through a perfectly selective semi-permeable membrane that blocks all non-water molecules. Given dilute conditions, the osmotic pressure (Π) across the membrane is determined by the Van't Hoff approximation to be proportional to the solute concentration in the reference solution: $\Pi = RTC$. Here, C refers to the sum of the total solute concentrations, accounting for anions, cations, and non-charged molecules (e.g. $[K^+]$, $[CO_2]_{aq}$, $[HCO_3^-]$, $[CO_3^{2-}]$).

The following known effects are neglected in this model: interactions of ions in solution, entropic contribution based on the differentiation between solute species, partitioning of DIC species as the system is concentrated and diluted, and any membrane-specific effects, such as concentration polarization[†] or ion-specific selectivity.^[28,46,47] Indeed, a more complete model would account for the thermodynamic activities of species in the electrolyte mixture and would include a solution-diffusion component for membrane effects;^[48] such analyses may be the subject of future studies.

The ideal RO work per mole of concentrated CO₂ for a reversible process with no dissipation is:

$$w_{RO,min} = RT(C_i/C_{out})\ln(\chi) \quad (13)$$

Where $C_i = A_i + C_{DIC}(A_i, p_i)$, χ is the concentration factor, and C_{out} is the concentration of outgassed CO₂ calculated from carbonate equilibrium assumptions (see SI Sections D.3.1 and D.3.2 for full derivation of the RO model). The logarithmic scaling with χ requires significantly less work at lower initial alkalinities (see Figure 4A). Physical systems approach this bound if the driving pressure is varied so as to be minimized throughout the entire concentration process.

A "single-stage" RO (ssRO) mode is driven by a single, fixed applied pressure throughout the concentrating process. Because we assume a perfectly selective membrane, the choice of the applied pressure is set only by the maximum concen-

tration in the concentrated state. A single-stage system is simpler to construct but has higher energy dissipation because the applied pressure is substantially greater than the counteracting osmotic pressure in the early phase of the concentrating process. The work per mole in the ssRO process is given by:

$$w_{\text{ssRO}} = \alpha_{\text{ss}} RT (C_i/C_{\text{out}}) (\chi - 1) \quad (14)$$

We include $\alpha_{\text{ss}} (\geq 1)$ as a scalable parameter to account for additional dissipation in physical RO systems; in the limiting case for ssRO, $\alpha_{\text{ss}} = 1$.

A “multi-stage” RO (msRO) process is made up of a series of ssRO modules. Instead of setting one driving pressure for the entire process, multiple driving pressures are chosen in order to reduce dissipation. If each ssRO subcomponent has an associated concentration factor of χ_{ss} , the work per mole of CO_2 is then:

$$w_{\text{msRO}} = \alpha_{\text{ss}} RT (C_i/C_{\text{out}}) (\chi_{\text{ss}} - 1) \log_{\chi_{\text{ss}}}(\chi) \quad (15)$$

Here χ is still the overall concentration factor of the entire msRO system. We use the log scaling as a simplification, even though physical systems would typically be constructed from discrete single-stage modules and thus would more closely be expressed mathematically through a summation series (See SI Section D.3.5).

In Figure 4A, we compare the energy cost of the ssRO, msRO and ideal models at a single representative input alkalinity value (10 mM) over a range of concentration factors, normalized by the α_{ss} parameter. In the ssRO case, energy cost rapidly increases with concentration factor. In contrast, the msRO model, at the same initial alkalinity, is significantly more energetically favorable at high concentration factors than ssRO. The results of the msRO model are reported in Figure 4B for various initial alkalinity values. The low initial alkalinity condition (1 mM) exhibits non-monotonic behavior, with a minimum around concentration factor of 100.

4.2.2 Technological implementation

RO technology can be implemented to drive the ACS by applying pressure to selectively pass water through a semi-permeable membrane, thereby concentrating the remaining solution. RO systems can operate across a wide range of con-

centrations, but have been most technologically tailored for seawater conditions, and tend to be tuned around producing a low-concentration, potable solution, rather than optimizing the concentrate parameters.

Typical seawater RO systems operate at around 80 bar and recover 50% by volume of the saline feed (roughly 0.6 M of NaCl equivalent or 35 g/L) as freshwater. For the purposes of the ACS, concentrating 0.6 M of input alkalinity by a factor of 2 outgasses 13 mM of CO_2 relative to the feed. We relate the energy cost of actual RO facilities to the theoretical ssRO work of 0.78 kWh per cubic meter of feed solution (typically reported as 1.56 kWh per cubic meter of freshwater), by evaluating the parameter α_{ss} .ⁱⁱ Current energies of medium to large capacity industrial seawater systems usually range from 1.1 to 1.25 kWh per cubic meter of feed (i.e., α_{ss} ranging between approximately 1.4 and 1.6),^[49] with newer facilities regularly achieving lower than 1.0 kWh per cubic meter of feed ($\alpha_{\text{ss}} < 1.3$).^[29]

Brackish water RO tends to operate at lower salinities, typically 5–200 mM of NaCl equivalent. For the ACS at a concentration factor of 10, an initial alkalinity of 10 mM or 100 mM yields 1.9 mM or 11 mM, respectively, of extracted CO_2 (Figure 2A). Currently deployed brackish water systems tend to be less efficient than seawater RO systems, although they have the capacity of having a much lower required energy — below 1 kWh per cubic meter of feed.^[50] For example, one industrial system that takes an input feed of 0.075 M NaCl equivalent with a concentration factor of 2 requires 0.445 kWh per cubic meter of feed ($\alpha_{\text{ss}} = 5.2$); another industrial system, which takes 0.063 M of NaCl equivalent feed with a concentration factor of 4 requires 0.825 kWh per cubic meter of feed ($\alpha_{\text{ss}} = 3.75$).^[51] Initial concentrations below 10 mM (such as 1 mM in Figure 4) are not considered in this analysis because α_{ss} values are not reliably reported at such dilute conditions.

Detailed modelling studies have looked at how much industrial brackish water systems can be further optimized through the use of energy recovery devices and by tuning operating conditions. One simulation study reports 0.48 kWh per cubic meter of feed, given 0.25 M NaCl equivalent feed and concentration factor of 2.5 ($\alpha_{\text{ss}} = 1.1$).^[52] Another detailed modelling study reports 0.30 kWh per cubic meter of feed, given 0.25 M NaCl equivalent feed and a concentration factor

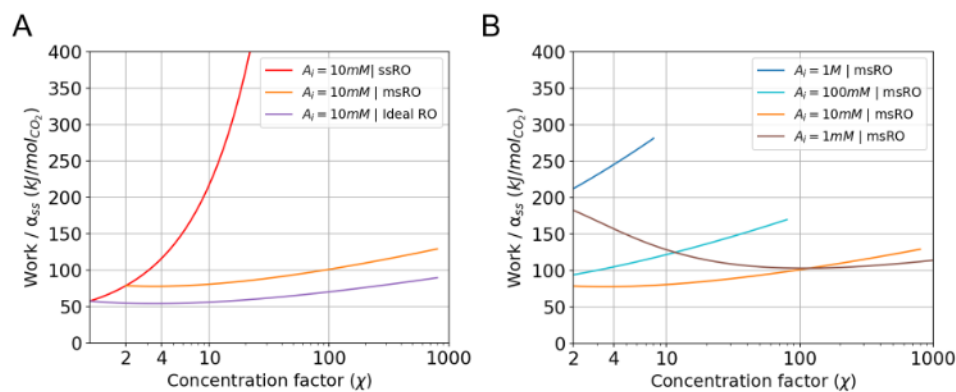


FIGURE 4 RO energy models. (A) Energy per mole CO₂ as a function of concentration factor for three RO models: ideal, multi-stage, and single-stage at a representative initial alkalinity of 10 mM. (B) Energy of the multi-stage RO model evaluated at four different initial alkalinity values. (A-B): energy is normalized to the scaling parameter α_{ss} ; curves terminate on the right at 8M precipitation threshold for potassium carbonate; for all curves $p_f = 0.4$ mbar; msRO model assumes ssRO subcomponent of $\chi_{ss} = 2$. Associated vacuum work (~ 30 kJ/mol) is not included; see Table 1 for estimate of full implementation energies.

of 5 ($\alpha_{ss} = 1.2$).^[53]

The significantly smaller values of α_{ss} from detailed modeling results compared to those for deployed industrial systems suggest that industrial brackish water systems can approach seawater systems in terms of energetic efficiency through straightforward modifications. The improvement in α_{ss} values is seen for systems with concentration factors ranging from about 2 to 5, suggesting that efficiency improvements could be applied to a wide range of system designs.

Although we rely on demonstrated RO systems to estimate the work required for driving the ACS, there are significant differences between ACS and desalination applications that must be considered. For example, brackish and seawater desalination must account for a complex variety of naturally occurring salt ion species and foulants present in seawater, which would not in general be the case in engineered ACS systems. On the other hand, RO membranes are not designed to be used for gas extraction and separating gases during the desalination process, which suggests further engineering modifications must be explored. Moreover, increased hydrostatic pressure as a result of the RO process modestly affects the carbonate equilibrium constants that are at the core of the ACS approach.

4.3 Ion binding-driven ACS

4.3.1 Energy model

Second, we pose a model for the work required to selectively remove water molecules from solution. This idealized “ion binding model” assumes that the energy to concentrate is dominated by enthalpic interactions, where a characteristic energy is associated with binding ions in solution, rather than entropic effects as assumed in the RO model. For a reversible ion adsorption process, the energy of binding ions from a feed solution and then releasing them into a concentrated stream sets the work necessary to concentrate the feed solution.

Here, we develop the simplest case consistent with experimental data:^[54] a constant electrical energy cost associated with binding an ion of a given charge out of the feed solution, independent of the concentration of ions found in the feed solution. We assume a value, ϵ_{ion} , for the energy cost to bind a pair composed of a monovalent anion and a monovalent cation, and we double that for a pair of divalent ions ($2\epsilon_{ion}$). This constant energy relationship may be observed when the selection mechanism applies charge or electric fields to do work on ions, rather than the uncharged water molecules of the solution, as in the RO energy model.

Such a model significantly simplifies physical effects as it neglects the following: ion-specific differences in binding

energy, increasing binding energy as a function of number of bound ions, additional energy cost or energy recovered from unbinding the ion, including the concentration of the solution into which the ion is unbound, and entropic and electrostatic effects of confining ions to different concentrations.^[47] In general, entropic factors imply that the work to bind ions should depend at least weakly on solution concentration, and electrostatic factors imply that binding energy per ion will increase above some density of bound ions. Additionally, divalent ions may have different binding energies than pairs of monovalent ions, due to both entropic and enthalpic effects.

Nonetheless, the application of this formulation to the ACS allows us to evaluate the associated scaling relation of electrical work given initial and final alkalinities and partial pressures of CO₂. The result of the model is that the required work, per unit volume, to concentrate the feed stream, assuming monovalent cations and bicarbonate and carbonate anions, is then directly proportional to the concentration of ion charges in solution. We write the binding energy per mole of ions by setting $\epsilon_m = \epsilon_{\text{ion}} N_A$, where N_A is Avogadro's number. The charge concentration is then set by the alkalinity, so the total binding energy per volume and per mole is given by $\epsilon_m A_1$. The work per mole of outgassed CO₂ is then (see SI Section D.4):

$$w_{\text{IB}} = \frac{\epsilon_m A_1}{C_{\text{out}}} \quad (16)$$

In Figure 5A shows the required work in the ion binding model per mole of CO₂ vs. initial alkalinity for various final alkalinity values at an outgassing pressure of 0.4 mbar (a constant vacuum energy value must be added to compare the total necessary work). This type of plot is more useful to assess the ion binding model than plotting work vs. concentration factor, as in Figure 4, because the physical and geometric properties of an ion-binding device are likely to set a constraint on the final alkalinity rather than the concentration factor. The minimum work per mole of outgassed CO₂ is reached at the limit in which the feed stream of DIC consists entirely of bicarbonate ions at low alkalinity. The "ideal limit" indicates the limit in which, at high alkalinity, all of the bicarbonate ions disproportionate to carbonate ions and CO₂ (Equation 8) and a maximum of 50% conversion is reached. At this limit the work is $2/\epsilon_m$ because two alkalinity carrier ions are bound

for each CO₂ molecule outgassed.

For any initial alkalinity, Figure 5 shows that the higher the final alkalinity, the higher is the C_{DIC} outgassed as CO₂ and the lower is the energy per mole of outgassed CO₂. In the context of this model, it is optimal to concentrate as much as physically possible as it does not penalize higher concentration factors. We note that, whereas energy efficiency is best for lower initial alkalinity values, the input stream C_{DIC} is also accordingly low, which means more water handling is required per mole of outgassed CO₂.

We show in Figure 5B that, for a given value of the final alkalinity, the total DIC outgassed as CO₂ as a function of initial alkalinity exhibits a peak. This occurs because higher initial alkalinities hold higher DIC but, as the initial alkalinity approaches the final alkalinity value, a smaller fraction of that DIC is converted to CO₂ and outgassed. This peak represents a further trade-off between outgassing concentration and outgassing energy built into the ACS as a result of the behavior of the carbonate system.

4.3.2 Technology implementation

CDI technology can be implemented to drive the ACS by using electric fields to do work on ions at approximately a constant energy cost per ion. CDI systems tend to operate best in or just below brackish water salinities, with salt concentrations typically in the 5–200 mM range.^[30]

Specifically, Zhao et al. showed experimentally that MCDI technology, which makes use of ion exchange membranes placed between the feed solution channel and the electrode, can operate at a value of ϵ_{ion} that is nearly independent of concentration. This occurs under constant current conditions over the entirety of the brackish water range, from 10–200 mM of NaCl.^[54] In this study, as in the ion binding model, this energy is also independent of the ion concentration in the concentrate stream. These results justify applying the ion binding model to ACS-CDI systems as a first-order study of energy scaling.

The values from MCDI studies are optimized given a condition on dilute stream purity, which, in the case of the ACS, unlike desalination, is not a relevant optimization target. In a limiting case, an excess of feed solution could be passed by the electrodes such that the dilute stream alkalinity is essentially the same as the feed. The potential to decrease ϵ_{ion} for

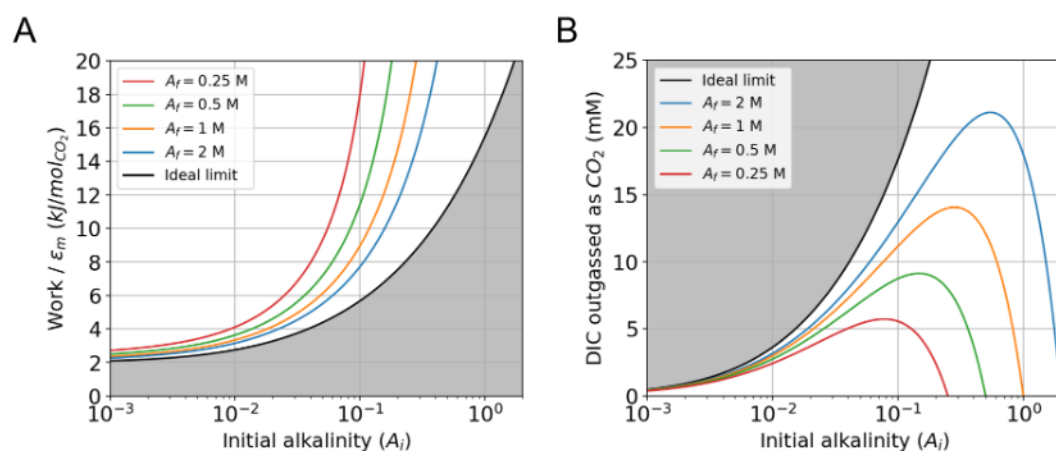


FIGURE 5 Results of ion binding model, calculated for various fixed initial alkalinity values. (A) Dependence on initial alkalinity of required work, normalized by ϵ_m to allow for rescaling to physical systems. Associated vacuum work is not included. In the ideal limit, the bicarbonate concentration fully disproportionates and half outgasses as CO₂. (B) Dependence on initial alkalinity of the concentration of DIC outgassed as CO₂. Curves cross 0 outgassed CO₂ at the point where the initial alkalinity equals the final alkalinity. In all cases the final outgassing pressure is set to 0.4 mbar. The ideal limit here is the same as for A.

differing solution optimization targets is an important subject for future studies.

Qin et al. describe existing CDI systems that are capable, for a ~30 mM salt feed stream, of recovering 95% of the water while rejecting 90% of the incoming salt,^[47] which corresponds to reaching a final alkalinity of ~500 mM. Suss et al. state that CDI can achieve 50% water recovery for sea water,^[30] which would correspond to a final alkalinity of ~1 M, and has the potential to achieve even higher water recovery ratios. This range of final alkalinities sets the regime explored in Figure 5, with the understanding that existing CDI systems, designed for desalination and not necessarily a high water recovery ratio, have not yet been optimized to produce a high-alkalinity concentrate stream.

An additional property of capacitive systems, which depends on operating parameters such as cycle rate, flow rate, and the current and voltage control, is that some amount of energy is able to be recovered as ions are released back into solution and current is reversed. As described above, the ion binding model does not account for this energy recovered from unbinding an ion, but it can be added to the model as an overall energy recovery factor. Energy recovery values for experimental systems operating at optimal conditions have

commonly been reported around 50%,^[54,47,55] with some studies approaching 80%.^[56]

4.4 Comparing estimated ACS implementation energies

Using the simplified models for reverse osmosis-driven and ion binding-driven ACS, we use parameters from the literature for implementations of RO and CDI, respectively, to estimate the energies per mole for different ACS processes if implemented with real physical systems. Here, we seek to estimate the energy that would be required to complete the full ACS cycle and extract CO₂ at 1 bar, so we include the additional work needed for vacuum outgassing, calculated in Section 4.1.

We caution that the values presented in this section are estimates that stem from plugging physical values into the simplified models above and do not represent a comprehensive prediction of the energy that will be required for these processes, at either industrial- or lab-scale. This will be the subject of future studies based on experimental work. The presentation of energies per mole, in Table 1, is to provide a basis for understanding over which ACS parameter ranges ei-

			ACS outgassing values			Energy estimates		
A_i (M)	A_f (M)	χ	C_{out} (mM)	f_{out}	$P_{t,max}$ (mbar)	RO (kJ/mol)	MCDI (kJ/mol)	MCDI w/ 50% Recovery (kJ/mol)
0.01	1	100	3.0	0.34	40	160-190	310-730	170-380
0.1	1	10	11	0.16	4.0	190-220	790-1900	410-970
0.6	1.2	2	13	0.039	0.80	250-310	—	—
1	4	4	31	0.055	1.6	350-420	—	—

TABLE 1 ACS outgassing values and implementation energy estimates for various initial and concentrated alkalinities. Here we assume $\alpha = 1.3 - 1.6$ and $c_m = 85 - 210$ kJ/mol, and we include the constant work to apply a vacuum at 0.4 mbar to be 30 kJ/mol. For the RO column, Row 3 corresponds to a ssRO model of $\chi = 2$ for approximate seawater conditions; all other rows are based on msRO using $\chi_{ss} = 2$.

ther RO or CDI may be more practical and over which ranges each might be practical at all.

For ACS-RO, we recall that the energy values for current seawater RO systems tend to be between 1.0 to 1.25 kWh per cubic meter of feed, corresponding to an α_{ss} range between approximately 1.3 and 1.6. We plug this range into the plots in Figure 4 for an initial alkalinity of 0.6 M, corresponding to seawater, and a concentration factor of two, corresponding to typical seawater RO, to obtain the energy estimates for RO in Table 1, Row 3.

In Rows 1 and 2 in Table 1, representing brackish water alkalinities, α_{ss} ranges from 1.3 to 5.2, the latter coming from the higher end of the range of α_{ss} values for deployed brackish water systems. We justify extending the lower range of α_{ss} down to 1.3 for lower initial alkalinities because of the simulation studies, cited in Section 4.2.2, that suggest industrial-scale facilities could be built with even lower α_{ss} values. The feasibility of deploying ACS-RO for brackish water with an α_{ss} of 1.3 is further substantiated in analysis by Qin et al. (2019), who show that the energy efficiency of existing brackish water RO deployments should not be taken as a limit because they have not yet reached the proper scale.^[47]

For ACS-CDI, we use the energies realized by MCDI systems because they can operate at a value of e_{ion} that is nearly independent of concentration and are also uniformly more energetically efficient than CDI without ion exchange membranes operated under the same conditions. MCDI ion removal energy values range from approximately 17 to 42 kT per ion, or 85-210 kJ per mol NaCl equivalent salt.^[30] One experimental study reports a value independent of input concentration of 22 kT per ion, or 110 kJ per mol NaCl equivalent salt, for an MCDI system operating over the range 10-200 mM NaCl.^[54] We use Equation 16 to obtain the energy estimates

for MCDI in Table 1.

As an example, we estimate the ACS-CDI energy for a particular set of parameters by applying the 110 kJ per mol value. If 10 mM input alkalinity is concentrated by a factor of 100 to 1 M, 3.0 mM of CO_2 would outgas at 330 kJ/mol. If we are able to achieve the commonly reported 50% energy recovery rate for the process, and then add the 30 kJ/mol for vacuum outgassing (Section 4.1), we would estimate the total work for the process to be 195 kJ/mol. That value falls in the range corresponding to these parameters in the Row 1 of Table 1, under "MCDI w/ 50% Recovery."

Finally, if we recalculate the ACS outgassing output and energy requirements using thermodynamic and physicochemical theories that account for ion-specific activity coefficients,^[57] we see a substantial enhancement to the results in Table 1, which is based on the same fixed equilibrium constant model we describe in Section 2. The shift in the DIC to alkalinity relationship due to the activity coefficient correction enhances C_{out} by 20-50% for each condition in Table 1, which reduces energy costs accordingly (see SI Section B.5).

5 Comparison to incumbent DAC technologies and feasibility of scale-up

In order for a particular DAC technology to contribute significantly to averting anthropogenic climate change, and contribute toward the gigatonne-per-year scale of global CDR that may be necessary by the end of the century, it needs to be able to be deployed at a large scale. To determine the feasibility of scale-up, we examine what challenges ACS implementations might face in reaching large-scale deployment levels and how

these compare to incumbent technologies, using the benchmark set by the National Academies of Sciences (NAS) report evaluating 1 MtCO₂/year DAC facilities.^[6] Whether through implementation using RO or CDI, the ACS possesses a number of potential advantages over incumbent DAC approaches but will also need to overcome some significant challenges.

5.1 Energy

The idealized energy requirement estimates provided for possible ACS-RO and ACS-CDI implementations in Section 4, while incomplete, can be compared to estimated requirements for incumbent DAC technologies. The lowest work from the conditions in Table 1 comes from Row 1, for which we estimate RO will require 160–190 kJ/mol and MCDI with recovery will require 170–380 kJ/mol. These estimates incorporate the work required to bring the outgassed CO₂ gas to 1 bar. As shown in SI Section B.4, this condition attains a purity of 99.8%. The NAS report estimates that the Carbon Engineering calcium loop-driven liquid solvent system has a work requirement of 360–480 kJ/mol (reported as 8.2–11 GJ/t) and that solid sorbent systems, for more realistic “mid-range scenarios,” have an energy requirement of 174–261 kJ/mol (reported as 3.95–5.92 GJ/t).^[6] These ranges are both for systems operating at a scale of 1 MtCO₂/year removed and for conditions comparable to our assumptions, capturing from a 400 ppm atmosphere at 25°C, with a 98% purity product.ⁱⁱⁱ

While the ACS idealized energy requirement ranges fall below that of the liquid solvent system and within the approximate range of solid sorbent systems, these ACS values should be viewed as far more uncertain when compared to ranges based on systems that have been realized at demonstration scale.^{iv} We note that the work estimate for MCDI can be improved if higher energy recovery factors, approaching the 80% factor reported for some systems,^[56] can be attained, though MCDI quickly becomes unfavorable at higher initial alkalinities when compared to both ACS-RO work requirements and to incumbent technologies. For ACS-RO, the work requirement increases less quickly for higher initial alkalinities and each of the other conditions in Table 1 remains in the range of the liquid solvent system.

Even without a final accounting of the energy requirements for implementing the ACS, we can compare the sources of

energy needed to incumbent DAC technologies. In the Carbon Engineering process, the core calcining step, for the final release of concentrated CO₂ from calcium carbonate precipitate, requires heating to ~900°C.^[6] Even though heat recovery is used for other processes that require low-grade heat and electrical energy is chosen when it can be used efficiently, the Carbon Engineering process still uses 5.25 GJ of natural gas per tonne of captured CO₂.^[22] Though the cost and energy requirements of the Carbon Engineering process account for offsetting of direct emissions, scaling up such a process would entrench an ongoing need for production of natural gas, as well as any emissions associated with the natural gas supply chain.^[58] Incumbent technologies relying on solid sorbents typically require heating to ~100°C during the desorption step of a thermal swing.

In contrast, the ACS does not require heat to operate, and can be operated entirely through renewable energy sources. Outside of the module that operates the ACS cycle itself, fluid pumping, for vacuum extraction and solvent pumping, can all be powered by electrical energy. An ACS-CDI module would require only electrical energy for ion binding,^[30] as would an ACS-RO module operating in the majority of the optimal regime described in Section 4.^[28]

5.2 Water

For any DAC approach that makes use of a liquid solvent for the initial capture of carbon dioxide from the air, it is important to determine the feasibility of the requirements for both overall water volume processing and water on hand. The ACS, as shown in Table 1, can use an incoming solution ranging from 10 mM to 1 M alkalinity, resulting in respective output concentrations ranging from 3.0 mM to 31 mM.

For 10 mM initial alkalinity solution, removing 3.0 mM each cycle means 7.6×10^9 m³ of solution volume needs to be processed to remove 1 MtCO₂ total. This is roughly an order of magnitude more solution than the annual processing rate for a large RO facility today, as we describe in Section 5.4. For 1 M initial alkalinity solution, removing 31 mM each cycle means 7.4×10^8 m³ of solution needs to be processed, decreasing the processing requirement to roughly that of a large RO facility.

The liquid solvent system of Carbon Engineering is the

key incumbent DAC technology that has significant water use. Carbon Engineering inputs 35,000 tonnes of a 0.45 M CO_3^{-2} and 2.0 M K^+ solution into its contactor per hour and uses it to capture 112 tonnes of CO_2 per hour,^[22] for a captured CO_2 concentration of 81 mM (using a density of 1.11 g/mL for the 2.0 M K^+ solution). Per unit of CO_2 removed, an ACS system would then require between 2.6 and 27 times as much water to be moved through the system each cycle.

The Carbon Engineering system requires an incoming solution stream of high alkalinity and high DIC. The ACS, however, takes a dilute incoming solution stream that is less constrained to a particular concentration. Air contact can thus be achieved with the additional surface area of the dilute solution, for example, with large pools for passive contacting. Large pools, with a high surface-area-to-volume ratio have significant water losses to evaporation as a function of humidity — these losses can be mitigated by either locating facilities in humid or rainy regions or by replenishing the pools. A dilute incoming solution stream also requires an increased energetic cost for liquid handling, although it reduces the energy requirement for operating contacting fans.

An ACS facility would require significant amounts of water to be stored on hand in pools near the facility due to the slow equilibration of the CO_2 -depleted solvent stream as it absorbs CO_2 before again being cycled through. These pools are where the feed stream for the ACS process would be drawn and where the combined concentrated and dilute streams would be returned, as shown in Figure 3A. For a $1 \text{ MtCO}_2/\text{year}$ facility, assuming large passive contacting pools of 0.1 meter depth, Stolaroff et al. provide a method for estimating the passive CO_2 uptake rate as a function of alkalinity.^[44,59] Using the ACS conditions, an instantaneous uptake rate of $5 \times 10^{-7} \text{ mol/s/m}^2$ of area for a 10 mM solution and $8 \times 10^{-7} \text{ mol/s/m}^2$ for a 1 M solution. If we use this as an estimate for the rate throughout the equilibration process, for the 3.0 mM and 31 mM CO_2 extraction quantities in Table 1 we obtain a characteristic equilibration timescale, τ , of 7 days and 50 days, respectively. For the 10 mM condition, this would mean keeping roughly 10^8 m^3 of water on hand in the pools, equivalent to the volume of approximately 40,000 Olympic swimming pools, cycled approximately 50 times per year.

These rates are applicable for the concentration of hydrox-

ide ions in the maximally CO_2 -depleted solution, however, and therefore represent the largest absorption rate over the duration of the equilibration process, resulting in a lower bound for the equilibration timescale. Assuming that this uptake rate slows approximately exponentially, for an elapsed time t we expect only the first $1 - e^{-t/\tau}$ of the equilibration process to have been completed. Over 7 days we would expect approximately 60% of the equilibration process to have completed. Cycling this solution through an ACS system would result in approximately 40% less CO_2 to be extracted for a similar energy cost. An engineering tradeoff between rate and energy cost would thus need optimization; e.g. waiting twice as long permits the equilibration to proceed to roughly 90% completion but requires keeping roughly $3 \times 10^8 \text{ m}^3$ of water on hand.

There is also a variety of ways to mitigate the extended timescales estimated above by introducing mixing into the contacting pools, instead of equilibrating in passive, unmixed pools. Depending on area constraints and the type of mixing used, this could be done either by continuously combining the outlet streams from the ACS system into one contacting pool, achieving a steady-state DIC concentration, or by using many smaller pools. Mixing can be done actively, as is done at some water treatment facilities,^[60] at an additional energetic cost. If we choose locations with high enough ground wind speeds, we can also expect the pools to remain well-mixed by the wind.^[61]

5.3 Land

As land continues to become a more limited resource, it is important to determine the land use requirement for any DAC technology. This land use requirement includes the footprint of the DAC facility itself (which, for the ACS, is predominated by the footprint of passive contacting pools) and the footprint of generation facilities required to power the DAC facility.

We estimate the land area requirement to power an ACS facility using only renewable energy sources, which requires substantially more land than using fossil energy, by using a result from Fthenakis and Kim that 1 Mha of land is required for 78.2 GW of solar power generation.^[62] Taking the upper end of the range for ACS-RO for a 10 mM solution, 190 kJ/mol (or 4.3 GJ/tCO_2), we then determine that

we would need 1.7×10^3 ha for the solar power to produce 1 MtCO₂/year. This land use requirement for power generation is much smaller than that for the contacting pools. From the estimate above in Section 5.2, to remove 1 MtCO₂/year with pools of 0.1 meter depth would require 10^9 m² (10^5 ha). For context, this is a large area — the Great Salt Lake (Utah, U.S.) is about four times the size.

We compare the ACS to incumbent technologies' land use requirement only for the facility itself, as the power generation footprint for each technology simply scales with the amount of power needed for a particular type of power source. Even when accounting for the indirect land impact, to ensure there are no detrimental environmental impacts when multiple facilities are built in the same area, the Carbon Engineering approach requires only approximately 700 ha to produce 1 MtCO₂/year, and much of this land could likely be used more flexibly; solid sorbent approaches require even less land.^[6]

To scale up the ACS to gigatonne-scale using passive contacting pools would require a likely unfeasibly large portion of land, on the order of 10% of the area of the US. Implementation of active mixing approaches, described in Section 5.2, could reduce this land requirement, though would carry an energetic cost. If mixing is used, the land use requirement can be reduced by an even larger factor than the water on hand requirement because, without the need for passive contacting, much deeper pools could be used.

5.4 Feasibility of scale-up and opportunities for optimization

An important advantage of the ACS approach to DAC is its ability to leverage existing technologies for water purification and desalination that are widely deployed at commercial scale around the world. Large RO facilities have capacities of more than 50 million m³ of purified water per year, with the largest plants having capacities of more than 350 million m³ of purified water per year.^[37] As described in Section 5.2, plants of this largest size implementing ACS (for example, using the condition from Table 1, Row 4 outgassing 31 mM of CO₂) would be able to capture up to 1 MtCO₂/year. Global desalination capacity is currently roughly 35 billion m³ of water per year and growing rapidly. So, achieving a scale on the order of 100 MtCO₂ captured per year seems feasible

based on current RO deployments, with larger scales achievable over time, though this would likely require substantial improvements in land use and water on hand requirements.

From the preceding sub-sections, we observe a clear trade-off between the required energy use, water on hand, and land use and the total water volume processing capacity to deploy a 1 MtCO₂/year ACS-DAC facility. In Table 1, we see the general trend that higher initial alkalinities have increasing energy requirements. On the one hand, whereas the 1 M initial alkalinity condition in Row 4 remains in the range of the incumbent liquid solvent DAC system for RO, this condition is far too energetically costly for CDI. On the other hand, we showed in Section 5.2 that the 10 mM initial alkalinity condition in Row 1 requires roughly an order of magnitude more water to be processed than the Row 4 condition, which itself already requires roughly the water processing of a large RO facility to achieve 1 MtCO₂/year. Without any improvements, then, the water processing scale for this low alkalinity Row 1 condition may be very difficult to achieve. Because its equilibration rate is more than twice as fast as the Row 2 condition, however, the low alkalinity Row 1 condition is favored with significantly lower water on hand and land use requirements. These trade-offs demonstrate both the challenges in scaling ACS-DAC for most conditions and the possibilities for tailoring the optimal ACS implementation.

Although contacting requirements pose challenging land and water demands, several mitigation options are possible to increase absorption kinetics and outgassing efficiency. Beyond the simple conditions examined in this study, different choices of solvents and membranes could enhance the ACS. Whereas only a strong base solvent was considered here, preliminary analysis shows that the right choice of a weak base solvent could increase the ACS outgassing amount (See SI Section C). In CDI systems the use of an ion-exchange membrane tuned to select for bicarbonate ions over carbonate ions could increase ACS efficiency. More generally, engineering membrane or electrode properties around bicarbonate and carbonate ions is an important area of future study. Beyond the two specific technologies explored in this analysis, the broad suite of existing desalination approaches,^[36] as well as hybrid approaches that combine strengths of different methods,^[63] could be investigated as driving mechanisms for the ACS. Furthermore, principles from the ACS could be explored as

a way of modifying and enhancing other solvent-based DAC methods. Finally, as described in Section 5.2, there are several ways to mix the contacting pools to increase the liquid-air contact area and increase the CO₂ absorption rate.

In addition to ensuring equitable allocation of scarce land and water resources, associated environmental hazards and material considerations must be considered before ACS systems are scaled up. Environmental impacts from traditional desalination facilities come predominantly from discharge of brine and from chemical treatment of water and membranes. Current membrane cleaning methods make use of toxic substances, which would need to be disposed of safely if large-scale systems are deployed.^[64] Because ACS systems can be operated in a closed cycle and, once operational, no significant discharge or uptake is necessary, these environmental harms can likely be mitigated.^[28] Although no significant loss of potassium or any other alkalinity carrier is expected, scaling up large alkaline pools would need to be accompanied with local environmental assessments to evaluate risk and determine mitigation options for spills or leakage.

6 Conclusion

The Alkalinity Concentration Swing is a new approach to DAC in which the driving mechanism is based on concentrating an alkaline solution that has absorbed atmospheric CO₂. Concentrating a solution with a given alkalinity and DIC results in disproportionation of bicarbonate ions into aqueous CO₂ and carbonate ions, proportionally increasing the outgassing partial pressure (Equation 11). This allows for extraction and compression of CO₂. For the same concentration factor, higher initial alkalinity solutions outgas a greater amount of CO₂ relative to the initial feed (Figure 2B). For a given final alkalinity, the amount of CO₂ outgassed vs. initial alkalinity exhibits a peak as initial alkalinity approaches that final alkalinity because of a trade-off between higher DIC available and a smaller conversion fraction of that DIC (Figure 5B).

The ACS can be implemented based on desalination technologies. We propose and briefly evaluate two technological implementation approaches, RO and CDI, with two accompanying simplified energy models, the RO model and the ion binding model, respectively. For each, the CO₂ capture energy

(Table 1) is dependent on the initial alkalinity, the concentration factor, and the applied vacuum pressure, as well as the dissipation for the associated implementation mechanism. We use reported experimental values from existing RO and CDI desalination implementations to estimate associated capture energies for ACS-RO and ACS-CDI. The choice of initial alkalinity and concentration factor present trade-offs between the quantity of outgassed DIC and energy requirements (Table 1). For example, a solution initialized at 10 mM alkalinity and concentrated by a factor of 100 outgasses approximately 3 mM of CO₂ and requires a lower bound of 160 and 170 kJ/mol for the msRO and MCDI (with energy recovery) models, respectively. A solution that swings between 1 M and 4 M alkalinity, however, outgasses 31 mM of CO₂, which corresponds to a factor of 10 less water processing, but requires a factor of about two more energy (lower bound of 350 kJ/mol, given the msRO model). Whereas the ACS idealized energy requirement ranges fall below that of the liquid solvent system and within the approximate range of solid sorbent systems, these ACS values should be viewed as far more uncertain when compared to ranges based on systems that have been realized at demonstration scale.

Overall, the ACS appears to be an intriguing DAC method in need of experimental research to further evaluate its viability. In particular, studies of ACS kinetics are important in order to understand possible limitations in the absorption and outgassing steps. Our analysis further reveals a trade-off for the ACS between the total water processing requirement and both the capture energy demand and water on hand requirement. Although initial calculations point to challenging land and water requirements to scale up this technology, we propose several potential mitigation pathways. Both ACS-RO and ACS-CDI approaches can be run entirely on electricity and do not rely on heat; the required materials are relatively simple (K⁺, membranes, electrodes, water). An initial assessment points to relatively environmentally safe deployment because no toxic chemicals, such as amines, are critical for this process. Lastly, the proposed approaches are based on existing technologies that have been deployed at large scale, and significant research and development can be leveraged from the desalination industry.

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Conflict of interest

There are no conflicts to declare.

Supporting Information

See associated Supporting Information at the following URL: <...to be updated...>

endnotes

- ⁱ Seawater RO processes account for the concentration polarization effect by running the maximum driving pressure roughly 10% higher than the value set by the direct osmotic difference.
- ⁱⁱ The ideal RO work is 0.53 kWh per cubic meter of feed solution (or 1.06 kWh per cubic meter of freshwater).^[29]
- ⁱⁱⁱ NAS report energy ranges are based on a CO₂ capture efficiency of 75%. We do not directly consider capture efficiency for the ACS, however, because ACS processes rely on passive contacting pools. Instead, as described in Section 5.2, we consider the timescale associated with the passive equilibration of those pools.
- ^{iv} For example, the energy of liquid pumping has been neglected in this analysis. If the scale of the additional energy cost per mole of CO₂ due to pumping is roughly approximated by the work needed to raise all processed water by 10 m then, for the conditions for which 3.0 mM and 31 mM of CO₂ is outgassed (Table 1), this would correspond to roughly an additional 30 and 3 kJ/mol, respectively.

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