

A Technology Development Matrix for Carbon Capture: Technology Status and R&D Gap Assessment

Ryan Baker¹*, Ommolbanin Alizadeh Sahraei^{1,2}, Mauro Mario Dal-Cin¹ and Farid Bensebaa¹

¹Engineering Division, Energy Mining and Environment, National Research Council of Canada, Ottawa, ON, Canada, ²Department of Chemical Engineering, Université Laval, Québec, QC, Canada

Numerous carbon capture, utilization, and storage (CCUS) technologies are under development to reduce CO₂ emissions. To evaluate the status of a CCUS technology under development and identify potential gaps for further advancement, we have established a new technology assessment framework and are developing a decisionmaking tool, the technology development matrix (TDM), starting with available carbon capture technology (CCT) data. TDM is a data inventory system and screening tool. As a screening tool, it can be used for resource allocation decisions in research, development, and deployment (RD&D) by academia, government, and industry. It shares data with techno-economic analysis (TEA) and life-cycle assessment (LCA) tools as an inventory system. By using available data, this TDM framework has been demonstrated on aminebased (monoethanolamine) absorption post-combustion CO₂ capture, for pulverized coal (PC) power plant flue gas, as the best available technology (BAT) for comparison. Three groups of promising post-combustion CCTs under development are presented as Alternative Technology (Alt Tech) case studies, including membrane, solid adsorption, and calcium-based chemical looping. By using available data, preliminary analysis enabled technology benchmarking and highlighted knowledge, data, and technology gaps, all providing potential future RD&D focus.

Keywords: carbon capture, decision-making tool, monoethanolamine, membrane, solid adsorption, calcium looping, techno-economic analysis, life-cycle assessment

1 INTRODUCTION AND CONTEXT

Carbon capture, utilization, and storage (CCUS) technologies are critical for achieving net emissions reductions (Haszeldine et al., 2018) and realizing the Paris Agreement Commitments (Board et al., 2019). However, due to the diverse characteristics of several components in the CCUS value chain, systematic planning for large-scale deployment of these technologies becomes a challenging task. The availability of various options for CO_2 emission sources (e.g., fossil fuel sources and industrial plants); capturing methods (i.e., pre-combustion, oxy-fuel combustion, chemical looping combustion, and post-combustion); utilization and storage technologies {including chemical synthesis [e.g., carbonates and cyanates]; energy products [e.g., methanol and hydrocarbons]; CO_2 injection [e.g., Enhanced Oil Recovery (EOR)]; oil and gas reservoirs; unmineable coal beds; deep saline aquifers; and CO_2 mineralization} can all be mentioned in this context. Furthermore, in integrated CCUS planning, to maximize economic profits while minimizing CO_2 emissions, several scalability issues (such as matching system components' time-based and

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*Correspondence:

Ryan Baker ryan.baker@nrc-cnrc.gc.ca

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Baker R, Alizadeh Sahraei O, Dal-Cin MM and Bensebaa F (2022) A Technology Development Matrix for Carbon Capture: Technology Status and R&D Gap Assessment. Front. Energy Res. 10:908658. doi: 10.3389/fenrg.2022.908658 spatial features, integration of CCUS into the energy grid, construction of pipeline infrastructure, coping with uncertainties in storage site's geophysical characteristics, social recognition, and legal framework) that can be addressed at various technological scales should also be considered (Tapia et al., 2018). Therefore, for the success of CCUS projects, there is a substantial need for developing planning frameworks and decision-making tools to help and facilitate identifying optimal solutions. The following paragraphs will review existing CCUS decision-making tools in the literature and give an overview of both commercial CCUS technologies, specifically Carbon Capture Technologies (CCTs), as well as promising ones under development. CCT for pulverized coal (PC) fired power plant flue gas was chosen because it has a complete data set.

First, a bibliometric analysis of CCUS literature reveals that, due to a lack of maturity, CCUS research is not as extensive as that of other low-carbon technologies such as solar and wind (Tapia et al., 2018). A few articles discussed how optimization tools can aid in CCUS planning. Tian et al. (2016) conducted a survey of process models to demonstrate how each component of CCUS can be modeled, highlighting unresolved issues for future progression in this field. A more recent review by Tapia et al. (2018) provided an in-depth overview of the advances in the planning of CCUS solutions, with a particular emphasis on process systems engineering methodologies that help in decision-making while taking multiple scalability issues into account. According to their survey, most CCUS planning tools used in the literature can be classified as mathematical programming, pinch-based analysis, automated targeting, and other computational approaches, such as numerical simulation and P-graph. While mathematical programming techniques have the advantage of simultaneously considering multiple factors that must be incorporated into a single model, they face some difficulties in communicating optimal model solutions to decision-makers, particularly when these solutions contradict common expectations (Tapia et al., 2018). On the contrary, pinch-based analyses require little computational effort because they are only capable of considering limiting factors at the same time, making finding solutions for larger problems difficult, whereas quick insights can be easily obtained for simplified or stylized problems by this method.

Generally, a comprehensive CCUS decision-making framework should mainly consist of i) energy models (focusing on an energy balance of CCUS systems), ii) pipeline infrastructure design focusing on a CO₂ distribution network), and iii) CO₂ source-sink matching, which can be further broken down into several subcategories (Huang et al., 2013). Attempting to address all these issues simultaneously by means of a single mathematical model can lead to over-complexity and unnecessary computational effort (Biegler and Grossmann, 2004). As an alternative, a multi-step approach can be adopted, with the results of each phase used to customize and simplify the subsequent step of model formulation (Tapia et al., 2018). In such a case, more accurate and realistic CCUS system planning can be accomplished for large-scale implementation through flexible and customizable tools created by the integration of individual component models. These individual assessment

tools that can be related to planning and decision-making for the CO_2 capture, transport, utilization, and storage components of a CCUS system should be able to be integrated to produce a single model. With this background, the present study focuses on establishing a technology assessment framework and developing a decision-making tool using available data to help experts develop/select an appropriate CCT as an important component of assessing a CCUS system's economic and environmental impacts.

Second, CCTs could be classified into two groups depending on where the carbon is captured: directly from the atmosphere or at the point of emission. The first group consists of direct air carbon capture, often referred to as DAC (Haszeldine et al., 2018). This technology is relatively new, with limited field data available in peer-reviewed literature. The second group consists of technologies enabling CO₂ capture at the point of emission from fossil fuel-based electricity production and industrial processes. In this second group, four CO₂ capture design approaches have been used for large-scale CO₂ emitters, including pre-combustion, post-combustion, oxy-fuel, and chemical looping combustion (Leung et al., 2014). Point emission, post-combustion, and chemical looping CCTs (the focus of this study) are the most widely developed, have available data, and are ready to deploy through retrofitting existing emitters.

Depending on the pre-capture and the required post-capture conditions and parameters, different groups of CO₂ capture approaches are available for post-combustion (Board et al., 2019), including cryogenic, absorption, membrane, adsorption, and chemical looping. Each of these CO₂ capture approaches consists of multiple process steps with options for optimization and hybrid approaches combining different technologies [such as membrane/liquefaction (Anantharaman et al., 2014) or membranes with scrubbing (Scholes et al., 2013; Freeman et al., 2014)]. When considering both diversity in CO₂ input characteristics (affected by the source of CO₂ emissions, such as a cement plant) and CO₂ output requirements (affected by the target CO₂ application, including sequestration), these different options give rise to a multitude of pathways at different research, development, and deployment (RD&D) Technology Readiness Levels (TRLs) of lab, pilot, and demonstration. In such a complex task, assessment frameworks and decision-making tools can substantially help technology developers in academia, government, and industry to recognize knowledge, data, and technology gaps to make better resource allocation decisions in order to accelerate RD&D of new CCTs and improve their competitiveness. Moreover, identification of the optimal CCT can also provide valuable feedback throughout RD&D in optimizing detailed engineering, techno-economic analysis (TEA), and lifecycle assessment (LCA), all using a shared data inventory.

After reviewing planning frameworks and decision-making tools, a few, if any, provide quantitative assessments based on specific data for comparing among different CCTs. The five-step sequence depicted in **Figure 1** summarizes the hierarchical process presented in this study to identify, assess, and optimize suitable CCTs among different options and find current knowledge, data, and technology gaps. Although early



	BAT	Alt Tech 1	Alt Tech 2	Alt Tech 3
CCT	MEA absorption	Membrane	Solid adsorption	Calcium looping
TRL	9	7	4–5	6–7
Description	Monoethanolamine, amine-based absorption	Polymeric membrane "Polaris"	Solid adsorbent, benzylamine pellet technology	Chemical looping, case 3 of calcium looping
Organization	US DOE, National Energy Technology Laboratory (NETL)	Membrane Technology Research (MTR)	ADA	South China University of Technology and Cornell University
References	Ciferno (2007)	Baker et al. (2018)	(Sjostrom et al. (2016); Stevenson and Armpriester (2016))	(Zhao et al. (2013); Tang and You (2018))

steps deal mostly with strategic and tactical decisions, later steps are more operational and less tactical. The preliminary strategic and tactical steps (stages 1 and 2, which are the focus of this study) enable comparing, ranking, and selecting among different technology candidates (mostly based on the input/output characteristic of CO_2), as well as identifying current knowledge, data, and technology gaps. This is done with available data, including modeling and simulation, lab experiments, and preliminary TEA and LCA studies. Detailed engineering, cost (TEA), and life-cycle (LCA) assessments (stages 3 and 4) help design and optimize the selected CCUS technology before construction (stage 5).

This study aims to establish a new technology assessment framework, develop a decision-making tool, namely, a Technology Development Matrix (TDM) (Malek and Nathwani, 2015), and apply it to CCT assessments. Excel is used to initiate TDM tool development; then, validation is done using available CCT data. As a research and development (R&D) decision-making tool and a data inventory system, TDM will help assess the current state of CCTs, benchmark them, and then help determine the next steps in ongoing TEA and LCA work (Zimmermann et al., 2018). Kearns et al. recently updated CCT-specific TRLs for various technologies, some specific to unique manufacturers (Kearns, 2021). Three different Alternative Technologies (Alt Techs) at different TRLs or phases of RD&D, including membrane, solid adsorption, and calcium looping, are investigated as case studies. Monoethanolamine (MEA) absorption was selected as the best available technology (BAT) because it is the only CCT currently demonstrated at a commercial stage. Hence, MEA is used in this work to

demonstrate the TDM's concept applied to CCTs and provide qualitative comparisons with CCT Alt Techs using radar plots. Details for the studied BAT and Alt Techs are summarized in **Table 1**. The TDM's quantitative comparisons, including technology ranking and benchmarking the three Alt Techs against the BAT, are the subject of ongoing work. Here, the BAT and three other Alt Techs are applied to flue gas from an operating, commercial PC-fired power plant as the base case or baseline. Knowledge, data, and technology gaps, as well as next steps, are summarized.

2 MATERIALS AND METHODS

In the following sections, the TDM framework's structure and overall approach (§2.1), the required steps to establish a TDM tool (§2.2), and TDM linkage to TEA and LCA (§2.3) are discussed in more detail.

2.1 TDM Framework Structure and Overall Approach

The TDM makes it possible to compare CCTs based on key criteria in three categories of performance, cost, and lifetime that can be measured through appropriate index metrics called key indicator parameters (KIPs). Performance, cost, and lifetime are broad categories of KIPs defined by technical operation, financial installation and operation, and years of operation, respectively. Three data sets of inputs for these KIPs are considered: the BAT, the user's Alt Tech, and a Target needed to be achieved. The Alt Tech is a CCT under development that the TDM user would like



to compare to both the BAT and Target. The Target is the set of KIPs that both the BAT and Alt Tech need to reach in order for R&D to achieve commercial deployment. Targets can be defined by a commercial market or application of the technology, as well as a goal defined by leading academic, industrial, or governmental organizations. The values of the KIPs for the BAT and Target can be affected by variables such as CO₂ target application and emission source, as well as the scale of capture technology. Therefore, in the first steps, these variables should be specified. The TDM is populated and designed in a way that the corresponding KIP data values for the BAT and Target will be retrieved from the database after the selection of the mentioned variables. Then, in the next step, the user will enter the KIP data values for the Alt Tech under study. The user can also change the TDM suggested KIP data values for the Target to a default value defined by users such as governmental organizations or R&D consortiums. In the absence of a Target for a particular KIP, the BAT KIP can be used as a guide.

The next stage is making qualitative and/or quantitative relative comparisons among BAT, Target, and Alt Tech using these three sets of KIP data. In a qualitative comparison (the scope of this study), a radar chart is generated that visualizes how close (in terms of KIPs) the user's technology (Alt Tech) is to the BAT and the Target. These results can be used to find R&D gaps and aid in decision-making for resource allocation. Regarding a quantitative technology ranking (the aim of our ongoing study), the ranking score of different user technologies will be calculated by analyzing and objectifying the relationships among the KIPs using a suitable method (e.g., combined scoring factor).

Figure 2 depicts the overall approach of the TDM, which is sorted by seven successive key blocks. This flow of information starts with selecting: i) the scale or size of the CCT, ii) the CO₂ emission source, and iii) the target CO₂ application. After selecting the above-mentioned variables, iv) the KIP data values for the BAT and Target will be retrieved by the TDM from its' database, and then. v) the user will enter the corresponding KIP data values for the Alt Tech under study. As mentioned before, in this step, the user also has the option to start with the default peer-defined Target values. The next step is related to vi) data processing by the TDM tool through qualitative/quantitative analysis of the three sets of data. Finally, vii) the results (radar charts/ranking scores) are produced, which could further be used for R&D decisionmaking purposes, as well as providing feedback for more detailed TEA/LCA analyses.

2.2 TDM Tool Development Steps

Creating a TDM tool entails five steps, as depicted in **Figure 3**, which will be discussed further in this section. For their initial development, the TDM tool and database were created using Excel.

2.2.1 Identifying KIPs

The first step is related to identifying appropriate metrics, allowing comparisons of CCTs based on three data categories of performance, cost, and lifetime, with all KIP data at the same TRL, and representing the entire CCT process being studied. These three data categories cover the main criteria required for



TABLE 2 | TDM KIPs for BAT, Target, and user's Alt Tech or studied CCT.

		Units ^a
KIPs	Performance Category	
	Scale of CC Process	tCO2/day
	CC Energy Efficiency	kWh/tCO2_cap
	CO₂ Recovery Efficiency	%CO₂ Rec
	Net Energy/tCO2 Emitted before CC	kWhe _{net} _wo CC/tCO ₂ _wo CC
	Net Energy/tCO2 Emitted after CC	kWhe _{net} _w CC/tCO ₂ _w CC
	HHV Efficiency before CC	%
	HHV Efficiency after CC	%
	% Loss Efficiency	%
	LCOE before CC	c/kWh
	LCOE Increase with CC	c/kWh
	% LCOE Increase with CC	%
	CO₂ Capture Cost	\$/tCO₂ Capt.
	CO ₂ Avoidance or Mitigation Cost	\$/tCO₂ Avoid.
	Cost Category	
	CCCL TCR	\$/tCO ₂ /day
	Variable O&M	\$/tCO ₂
	Fixed O&M	\$/tCO ₂
	Tech Replacement Costs	\$/tCO2
	Lifetime Category	
	Maximum Plant Life	Years
	Levelization Period	Years
	Key Tech Lifetime	Years

^at indicates a metric ton or tonne.

technology commercialization. As far as specific KIPs in each of these three categories, we are following the approach developed by the University of Michigan's Global CO₂ Initiative (GCI) (Zimmermann et al., 2018) and using available CCT data. Data categories of performance, cost, and lifetime were identified in the context of TDM development for grid scale, Front of the Meter Electrical Energy Storage (Malek and Nathwani, 2015; Malek, 2019; Shi, 2020), and National Research Council (NRC) of Canada's previous TDM work in Proton Exchange Membrane Fuel Cell (PEMFC) R&D. These three categories are further broken down into individual KIPs, measuring the CCT under study. The question in this step is, "What are the best representative KIPs for each category?" These parameters should cover plant energy efficiency before and after CO₂ capture (e.g., lower heating value (LHV) efficiency, higher heating value (HHV) efficiency, CO₂ capture penalty), CO₂ capture efficiency and capacity, investment effort [e.g., capital

expenditure (CapEx) including Total Capital Required (TCR)], processing effort [e.g., operational expenditure (OpEx)], cost of CO₂ (captured and avoided), electricity costs (Levelized Cost of Electricity (LCOE) before and increase with CC), and plant technical and economic lifetimes. **Table 2** shows an example of the main TDM data table with detailed KIPs for each category of performance, cost, and lifetime.

All masses are listed in metric tonnes of CO_2 or tCO_2 . Because available data are published in different years, the largest cost component—Carbon Capture Compression and Liquefaction (CCCL) TCR—is inflation adjusted from the costing year indicated in the publication to the most recent complete calendar year, which is 2020 at the time of this publication. Currencies are US dollars (\$) or US cents (c).

A sub-level of technology-specific KIPs can be added to make comparisons within one CCT. Future work could explore comparisons among one CCT and include technologyspecific KIPs.

2.2.2 Data Standardization

Data are critical for proper technology evaluation. A lack of standardized and systematic assessment methods prevents an accurate technology evaluation and makes benchmarking quite challenging. Newly detailed frameworks for systematic TEA and LCA of CO₂ capture and CCUS technologies have been published (Rubin et al., 2013b; Zimmermann et al., 2018). The work was coordinated by the GCI and supported by Europe's EIT Climate-KIC (Zimmermann et al., 2020a). This work did not assess the economic and environmental impact of different capture technologies.

A reliable and meaningful assessment of different pathways cannot be achieved without a similar (if not the same) project scope, system boundaries, KIP calculation methods, and technology maturity (Rubin et al., 2013b; Zimmermann et al., 2020b). Furthermore, a reasonable overlap between temporal and affecting conditions, geographical the corresponding assumptions and design parameters, is also required (Zimmermann et al., 2020b). Therefore, data from various studies must be finally reviewed and harmonized to ensure that the same units, calculation methods, cost year (i.e., accounting for inflation), CO2 compression pressure and purity, and plant scales are used (Kuramochi et al., 2012).

2.2.3 Identifying Options for Technology Scale, CO_2 Emission Source, and CO_2 Application

As seen in Figure 2, options for technology scale include laboratory and bench (Research: TRL of 1-3), pilot (Development: TRL of 4-6), and finally demo/commercial (Deployment: TRL of 7-9) scales (Buchner et al., 2018; Zimmermann et al., 2018). Laboratory and bench scales can be gram production of novel carbon capture (CC) materials and would serve to determine the disruptive potential of new absorbents, membranes, or adsorbents. For TRL 1-3, a simplification of the entire process can be used. Starting at TRL 4, all necessary equipment for that same process is included. This is the main CCT unit plus the Balance of Plant (BoP) or Balance of System (BoS). Pilot plants can be anywhere from <1% to 10% of the intended scale of application and would encompass all ancillary processes. The key technology from the laboratory and bench would be developed further to validate the concept and start scale-up preparation. At TRL 5-6, there would be well-established process diagrams, heat and energy balances, and process simulations. Finally, at TRL 7-9, demo/commercial can also vary widely, where conventional coal-fired power plants range from 200 to 600 MW, with the accompanying range of CO₂ emissions. The options for CO2 sources can be found in §2 in the Supplementary Material.

2.2.4 Building the TDM Database for BAT and Target In order to create the TDM's database of KIP values for the BAT and Target, data can be collected from complete TEA and LCA studies (at the same TRL) in available literature and reports based on the experimental or operating plant and process simulation results. In this work, published CCT data were used to create an Excel database. For details, see §3 in the **Supplementary Material**.

2.2.5 Constructing Data Analysis Model and Radar Plot

Qualitative radar plot outputs compare the user's Alt Tech, the BAT, and an aggregate of the Target application or goal. Radar plot outputs can be used by technology developers, such as academia, government, and industry (including small-tomedium-sized enterprises). Here, Excel was used to construct the initial data analysis model and radar plot.

For each KIP, the Alt Tech is first compared to the BAT, as shown in **Eq. 1**. Then, BAT and Alt Tech are compared to Target, as shown in **Eqs 2**, **3**, respectively. Using Cost CO_2 Captured ($/ \$ as an example KIP,

$$\frac{Alt \, Tech}{BAT} = \frac{Alt \, Tech \, Cost \, CO_2 \, Captured}{BAT \, Cost \, CO_2 \, Captured} \tag{1}$$

$$\frac{BAT}{Target} = \frac{BAT \operatorname{Cost} \operatorname{CO}_2 \operatorname{Captured}}{Target \operatorname{Cost} \operatorname{CO}_2 \operatorname{Captured}}$$
(2)

$$\frac{Alt \, Tech}{Target} = \frac{Alt \, Tech \, Cost \, CO_2 \, Captured}{Target \, Cost \, CO_2 \, Captured}$$
(3)

Different quantitative methods can be used for analyzing and objectifying the relationships among the KIPs (e.g., combined scoring factor) (ES-SelectTM, 2012). However,

these quantitative assessments require more data than available at the time of this report and are thus the subject of ongoing work. Quantitative methods include the analytical hierarchy process, mathematical programming, data envelopment analysis, and fuzzy logic.

2.3 TDM Linkage to TEA and LCA

TEA and LCA tools are used to assess CCUS technologies based on technical, economic, and environmental performance indicators. TEA is often performed to assess the economic (e.g., cost) impact of technologies identified based on a set of technical criteria along with CAPEX and OPEX data. Often done separately, LCA is used to assess environmental impact where energy efficiency and greenhouse gas (GHG) emissions are the most widely reported environmental indicators for those identified technologies. The inputs and outputs from these two tools are key in building the proposed TDM components (KIPs) required to provide preliminary technology screening. Separately, the CCT under study can be modeled to extract mass and energy flows, equipment, and operating and maintenance requirements for a specific scale to provide the necessary input data for TEA/LCA tools. Moreover, data can also be obtained from experiments and operating plants. After the initial screening of the CCTs by the TDM, in the next step, the potential successful deployment of the down-selected CCT(s) could be assessed by detailed TEA and LCA. As shown in Figure 4, in this way, the TDM tool, along with TEA and LCA tools, integrates within an overall framework enabling preliminary screening and final technology benchmarking. For the initial Excel TDM tool, data are exchanged manually with TEA and LCA tools. However, NRC has also developed an LCA and life-cycle inventory-based webserver, the Economic and Environmental Evaluation of Carbon Conversion (NRC, 2022). Future versions of the TDM tool will be integrated into (thus be able to share data with) NRC's webserver using a more automated process.

It is worth mentioning that, to exchange data among TEA, LCA, and the corresponding TDM, they must be standardized as outlined in §2.2.2 (Zimmermann et al., 2018). In this way, data sharing could result in accurate and better decision-making as technology progresses in scale and TRL from lab or bench top to pilot, demonstration, and finally commercial deployment.

The current embodiment of the TDM does not account for CO2 credits/debits based on LCA or sales of CO2 or reagents, such as spent CaO in calcium looping. Tang and You performed both TEA and LCA for different calcium looping scenarios but did not add CO2 accounting from the LCA analysis to the metrics calculated for the CO₂ capture process (Tang and You, 2018). Just as the CO₂ avoided cost is usually higher than the CO₂ capture cost because the former accounts for additional CO₂ produced and changes in power output for the CCT, LCA would assign additional CO₂ penalties to a CCT. These are all within the system boundary and can only be completely accounted for in the second and higher iterations of the process in Figure 4. However, other CO₂ penalties could be accounted for in the first iteration. Examples could be CO2 emissions associated with MEA production and breakdown products or mining of CaO/CaCO₃.



3 RESULTS

Three examples of alternative CC technologies, namely, membrane, solid adsorption, and calcium looping, all using flue gas from a baseline PC-fired power plant, are investigated in this section. Amine-based (MEA) absorption scrubbing retrofitted to a PC-fired power plant is considered the BAT (Ciferno, 2007) to assess the feasibility of Alt Techs. The three aforementioned promising CCTs are assessed using a qualitative radar plot in this study and compared to MEA.

The basic principles are provided in §3.1. Then, the selected MEA absorption as BAT is described in §3.2. An overview of the technological fundamentals of each post-combustion CCT followed by identification of the specific characteristics of the selected Alt Tech in that category is discussed in §3.3. Finally, the knowledge and data gaps in applying TDM to the selected Alt Techs are discussed in §3.4.

3.1 Baseline System Boundary

Generally, in a conventional coal-fired power plant, steam (which is generated through burning fossil fuel with air in a boiler) is transformed into electricity using a steam turbine. In the next steps, before emitting the flue gases into the atmosphere, NO_x , particulate matter (PM), mercury, and SO_x are removed. While in a retrofitted power plant with CCT, the flue gas, which mainly consists of N_2 , is emitted into the atmosphere after CO_2 capture. Therefore, in a retrofitted power plant, more resources (requiring additional energy and materials) must be used to enable CO_2 capture. This includes material treatment and circulation, heat exchangers, solvent/sorbent regeneration, and CO_2 purification and compression (Ciferno, 2007).

As mentioned in \$2.2.2, a meaningful and accurate comparison among different CCTs that can be used for retrofitting a CO₂ emission source plant (e.g., a PC-fired

power plant) cannot be achieved unless sufficient alignment in the overall system boundaries, major assumptions, and KIP measurement methods has been considered. A commonly used system boundary for assessing a retrofitted power plant includes the main constituents and auxiliaries from flue gas desulfurization (FGD) to CO₂ conditioning (CO₂ purification and compression). This means that, in addition to the CO_2 capture unit, the required flue gas pre-treatment, in particular expansion of the FGD unit, if required, and the changes to the boilers are also considered. Furthermore, purity, temperature, and pressure of the final compressed CO₂ product, CO₂ capture rate, year of cost estimate, plant size (gross power output), plant capacity factor, type of fuel used in the power plant, and conversions between LHV and HHV are some examples of major assumptions that should be suitably aligned and harmonized for a proper technology assessment. CO2 storage and maintenance charges are recommended but are normally not applied in most studies (Rubin et al., 2013a).

3.2 MEA Absorption as the BAT

Amine-based absorption technology is the most mature and commercially available process to capture and separate CO_2 from flue gas; as such, it is often used as a benchmark (Alstom UK, 2011; Anantharaman et al., 2016). MEA is the most widely deployed amine-based absorption technology. Although MEA is presently the most cost-effective CCT, it has drawbacks. MEA incurs a considerable energy penalty during solvent regeneration, has a narrow operating temperature range, and produces substantial waste during cycling, and the spent materials are difficult to dispose of (US DOE/NETL, 2015a). Currently, research is focusing on developing new amines and mixtures to improve energy efficiency and thus cost and reduce the overall carbon footprint.



Other potential scrubbers exist and are used on large scales. Selexol (dimethyl ethers of polyethylene glycol), Purisol (n-methyl-2-pyrrolidone), and Rectisol (methanol) are viable alternatives. These are used to remove CO₂ from natural gas wellheads or where H₂S is involved because H₂S can be removed from CO₂. Physical solvents, Selexol and Purisol, require high pressure and high CO₂ concentrations to achieve reasonable CO₂ loading. As a result, they are generally applied to CO₂ and H₂S removal in integrated gas combined cycle (IGCC) power plants (Chen, 2022). Rectisol is ideal for processes where gases are already at the right conditions for methanol to solubilize/ absorb H₂S and CO₂ (Hochgesand, 1970). Methanol spontaneously releases CO2 when pressure is released and avoids the parasitic power losses associated with MEA. These scrubbers are better suited to IGCC power plants and natural gas well heads than PC power plants.

The NETL study (Ciferno, 2007) of the Conesville #5 unit in Ohio is a rare example of studies reported in great detail. The Conesville #5 unit, shown in **Figure 5**, is a 460 MWe gross PC power plant producing 9,440 t_{CO2} /day. The study considers preexisting flue gas treatment and its expansion to meet MEA requirements, physical restrictions at the site, changes to boilers, waste disposal, and other factors. The TDM parameters (KIPs) relevant for this process as the BAT are shown in **Table 4**.

3.3 Alternative Carbon Capture Technology Case Studies

MEA is a relatively mature CCT with numerous large-scale demonstration plants. However, limited field data on alternative CCTs are available. Often, these Alt Techs are at lower TRLs, and any available data come from lab and occasionally pilot scales, combined with data from process simulations and TEA studies. As a result, there are limited Alt Tech data from operational units that include the entire process and use the same KIPs across all CCTs. Three promising CCT Alt Techs are presented below, for which there are enough data from lab and pilot scales combined with simulations and TEA studies. In the case where KIPs were not the same, the authors recalculated KIP values from the data.

3.3.1 Alt Tech 1: Membrane Technology

Although not yet commercialized but operated as large-scale pilots, 20 tCO₂/day and 200 tCO₂/day planned, membrane technologies are a relatively mature technology combining attributes related to scalability and potential future improvements (Merkel, 2018). The Polaris membrane developed by MTR is undoubtedly the most advanced system for CO₂ capture (Merkel et al., 2010; Baker et al., 2018). Membrane development consists of several stage gates, including the following:

- Development of a material for the actual CO₂/N₂ separation,
- Preparation of this material in a form suitable for large-scale production while achieving high throughput and defect-free production,
- Implementation of the membrane in an appropriate pressure vessel (module) and addressing the mass transfer of CO₂ and N₂ in the membrane module (Merkel et al., 2009),
- Optimization of the process flow diagram using these membranes (Wei et al., 2015),
- Demonstration of long-term flux and separation performance (Kniep et al., 2018).

The Polaris[™] membrane is a multilayer thin-film composite (TFC) prepared in flat sheets and enclosed in spiral-wound modules to separate feed and CO₂-rich product streams. The separation is affected at the top layer, which has a low permeability, but the high flux is achieved via the ultrathin thickness of this layer, under 100 nm. The MTR design is like the more familiar spiral wound modules used for reverse osmosis desalination of seawater but with critical modifications to the permeate side to avoid CO₂ build-up and allow for a "sweep gas," which dramatically improves the overall CC performance of Merkel et al. (2010). MTRs approach to improving its technology is toward developing higher throughput rather than selectivity. The first generation Polaris[™] membrane (Gen-1) achieved 1,000 gas permeation units (GPU) and a CO₂/N₂ selectivity of 50. Gen-2 membranes with a permeance of 2,000 GPU have been fabricated in prototype modules. Gen-3 membranes have achieved 3,000 GPU at the lab scale, with all three generations maintaining the initial selectivity (Kniep et al., 2018). Further improvements to the module design have reduced pressure drops and are projected to reduce power consumption by 15 MWe on a full-scale plant (Merkel, 2018).

Other approaches to ultra-high permeance membranes are achieved by using materials with higher permeability and thicker separation layers (Shao et al., 2013). Recent developments have seen further dramatic improvements, where CO_2/N_2 selectivity increases up to 242, with 823 GPU achieved (Wang et al., 2020). Further increases in permeance to 3,000 GPU, with an excellent selectivity of 78, have also been obtained (Qiao et al., 2019). Both of the latter membranes are still at the lab scale but have been prepared in large areas (100 cm²) for this stage of development, using polymers with reduced defects. While still at an early development stage, improvement in membrane quality is expected to be integrated into higher TRL CCT systems.

Other dramatic improvements in membrane technology were reported by McNeil et al. (2020). In this novel design, silver-based dual-phase ceramic-based membranes are used, giving rise to the selectivity of 1,000 and permeability of >200,000 Barrer. This class of membranes has great potential, as explained Anantharaman et al. (2014) in process simulations with IGCC and coal gasification-based power generation. However, these novel membranes are probably less suited to conventional coalfired plants. The dual-phase membranes require >600°C operating temperatures and, like Rectisol processes, can take advantage of process conditions to permeate CO2 and retain combustible gases at high pressure. Cost and scale up to areas required for CCUS are to be overcome, and the TRL for these membranes must be considered low, 1-2, may not advance quickly, and likely never reach the costs of polymeric TFC membranes (Merkel et al., 2010). However, they may be the only CCT option for IGCC.

Hollow fiber membranes are usually the design of choice because the membrane area/volume is typically 10 times that of spiral wound modules. This benefit of hollow fiber packing density was negated when membranes for CO_2/N_2 separations saw 10–100-time increase in flux with the PolarisTM membrane. Hollow fiber modules with such high GPU ratings have yet to be

produced from any polymer. Furthermore, concentration polarization (poor mixing on the feed side of hollow fibers on the shell side) is exacerbated with high flux membranes, and 1,000 GPU membranes (now 2,000 and potentially 3,000 GPU membranes) with a CO_2/N_2 selectivity of 50 would not see the theoretical performance in practice.

The MTR studies indicated that improved economics lie in the direction of increased flux after a given selectivity is obtained. This finding is specific to their approach, where low feed pressure and partial vacuum on the permeate side of the membrane are used as driving forces for CO_2 permeation. The study by Anantharaman et al. (2014) benefited from higher selectivity, but only at the very high operating pressures in that particular application, those pressures in a conventional PC post-combustion capture scenario would consume too much power.

This initial MTR study for 90% capture and liquefaction of CO2 from a 600 MW coal-fired plant (producing 11,040 t_{CO2}/ day) was later updated for a cost-optimized case as 80% overall capture (Baker et al., 2018) and used in this work. At the time of publication, Dillon reported on a cost estimate for retrofitting MTR's second-generation Polaris[™] membrane based on a DOE field study of an existing coal-fired power plant (Dillon, 2021). The basic block diagram of the MTR process is shown in Figure 6. A key process modification is using a sweep gas on the membrane stage before the stack, which returns a portion of the CO₂ to the boiler. This increases the overall flux and CO₂ concentration in permeates of the other membrane stages, thereby reducing CAPEX and OPEX. Similar approaches are used with natural gas turbine power plants, where the exhaust gas recirculation is used to increase the exiting CO₂ concentration for all CCTs, not only membranes.

3.3.2 Alt Tech 2: Solid Adsorption

CO₂ capture using solid adsorbents is another method in the category of post-combustion CCTs, which provides several advantages over the traditional commonly used liquid scrubbing (such as aqueous MEA absorbents). Compared to an aqueous solvent-based process, considerable reduction in the regeneration energy penalty (due to the solid adsorbents' lower specific heat capacity and decreased water evaporation), a wider range of operating temperatures (from ambient temperatures to 700°C), less waste production during cycling, and easier disposal of spent materials, are advantages of solid adsorbent-based technology (US DOE/NETL, 2015a). However, for large-scale CO₂ capture, in many or most cases, the adsorbent-based technology is still not as cost-effective as the solvent or membrane alternatives. The high cost of solid adsorbents [e.g., zeolites and Metal-Organic Frameworks (MOFs)] and adsorbent attrition are examples of the main challenges associated with this technology (US DOE/NETL, 2015a).

Solid adsorbents for CO_2 capture are typically categorized based on the operating temperature regimes for adsorption and desorption (regeneration). Indeed, they fall into three categories of i) low temperature (<200°C, such as supported solid amine-based, zeolite-based, MOF-based, carbon-based, and alkali metal



carbonate-based); ii) intermediate temperature [200°C-400°C, such as layered double hydroxide- (LDH-) based]; and high temperature (>400°C, such as CaO-based and alkali ceramic-based) solid adsorbents (US DOE/NETL, 2015a).

The performance of solid adsorbents can be evaluated based on different attributes such as their equilibrium CO_2 adsorption capacity, adsorption/desorption kinetics, CO_2 selectivity, multicycle durability (affected by their tendency for moisture uptake, tolerance to impurities, and mechanical strength), and regeneration (Samanta et al., 2012; Krutka et al., 2013).

Several research groups have investigated this subject in recent years, with many of them focusing on new material development (Samanta et al., 2012). However, to commercially use solid adsorbents for CO₂ capture, research on the development of the effective process, including gas-solid contactors (e.g., fixed-bed, fluidized-bed, and moving-bed), adsorbent regeneration process, and the overall process integration (primarily thermal integration with the emission source plant) is also required. In order to meet the technical and economic constraints, it is expected that researchers' focus will be more on investigating the use of innovative gas-solid contactor configurations (e.g., structured bed, fast-fluidizedbed, circulating bed, and transport-bed) (Samanta et al., 2012), adsorbent performance under actual operating conditions, reducing solid adsorbents' cost and operability issues (such as adsorbent degradation and associated emissions), and performing more precise and reliable process design and economics analysis (US DOE/NETL, 2015a; Pardakhti et al., 2019).

NRG Energy, Inc. (NRG), in collaboration with Svante (formerly Inventys), using Svante's VeloxoThermTM CCT (Stevenson and Armpriester, 2016), evaluated their solid adsorbent at a scale of 10 MWe. Their target was 90% CO₂ capture efficiency with a purity greater than 95% (Target defined by a solicitation (DE-FOA-0001190) from the U.S. DOE) (US DOE/NETL, 2015b). The VeloxoThermTM uses a

rotary adsorption machine, an evolution of well-developed, large-scale, rotary air preheaters. This technology relies on three key innovations: i) an intensified rapid cycle temperature swing adsorption (TSA) process, ii) a patented architecture of structured solid adsorbents, and iii) a rotary embodiment (Rezaei and Webley, 2009; Stevenson and Armpriester, 2016). Important advantages compared to traditional TSA processes are a low-pressure drop, immobilized adsorbent (eliminating mechanical degradation), high heat and mass transfer, and high heat retention. The NRG study identified shortfalls in the performance of the structured bed solid adsorbents, leading to reduced capacity and increased steam consumption. Recently, Hovington et al. reported the development of a zinc triazolate MOF adsorbent at large production volumes (Hovington et al., 2021) and demonstrated the technology on cement flue gas (Lin et al., 2021). The new adsorbent's properties surpass those of the ideal material in the NRG study, but the TEA for PC power plants has not yet been revised.

The study by ADA (Sjostrom et al., 2016), which is used as the Alt Tech for solid adsorption in the current study, evaluated benzylamine immobilized on pellets. The solid adsorbents were benchmarked against MEA, and two costing methods were presented, based on DOE guidelines and ADA best experience, where the latter TCR was 32% higher than DOE guidelines. The process flow diagram is shown in Figure 7. A preliminary plant design and TEA were based on lab-scale performance for a power plant with a net electrical output of 550 MW before and after CC. The adsorbent capacity for the initial TEA was 15 $g_{\rm CO2}/100~g_{adsorbent}$ and a contact time of 10–15 min, requiring four parallel adsorption/desorption trains. Pilot studies (Sjostrom et al., 2016) at a scale of 1 MWe were performed and showed reduced CO₂ capture performance with a reduced adsorbent capacity of 5.4 $g_{CO2}/100$ $g_{adsorbent}$ and an increased contact time of up to 65 min. The TEA study was repeated based on the downrated pilot plant performance: the



number of parallel adsorption/desorption trains increased from 4 to 44. The TDM in our work uses costing data based on the ideal adsorbent performance and the DOE guidelines. Both ADA and Svante studies had promising TEA results based on lab-scale data, but performance decreased at the pilot scale. Kearns et al. ranked the new solid adsorbent TRL from Svante at 5–6 in a CO_2 from capture scenario from cement production. As a result, we ranked the TRL for ADA's Solid adsorption at 4. However, future improvements in ADA's performance at the pilot scale are possible; if so, the TRL would increase.

3.3.3 Alt Tech 3: Calcium Looping

 $CaO/CaCO_3$ (or calcium) looping technology emerged over a decade ago as one of the promising technology pathways to achieve low-cost post-combustion CO_2 capture in the near future. In this process, the reversible reaction of calcium between its carbonate and oxide forms allows CO_2 separation from flue gases coming from either power plants or any other industrial CO_2 emission sources as follows:

$$CaO(s) + CO_{2}(g) \xrightarrow{Carbonation \sim 650^{\circ}C} CaCO_{3}(s) \Delta H_{298K}^{r} = -179 \, kJmol^{-1}$$
(4)

$$CaCO_{3}(s) \xrightarrow{Calcination \sim 850-950^{\circ}C} CaO(s) + CO_{2}(g) \Delta H^{r}_{298K} = +179 \, kJmol^{-1}$$
(5)

Shimizu et al. were the first to propose the use of a hightemperature CaO sorbent for post-combustion CO₂ capture, and it has since become widely used in calcium looping process research (Shimizu et al., 1999). A typical process flow diagram abstracted from the work of several active groups in this field is shown in Figure 8 (Zhao et al., 2013). Flue gas containing a low-to-medium concentration of CO₂ is fed to the circulating fluidized bed (CFB) carbonator, containing refreshed solid CaO sorbents, where the exothermic carbonation reaction (Eq. 4) happens at around 650°C. The CaCO₃ formed is then passed to a CFB calciner where the endothermic calcination reaction (Eq. 5) occurs around 850°C-950°C, resulting in regeneration of CaO (which is passed back to the carbonator) and the formation of an almost pure stream of CO2. The CO2-depleted flue gas is emitted into the atmosphere, whereas the CO2-rich stream is released for further purification and compression before being utilized or stored. Due to temperature differences between the carbonation and calcination reactions (650°C vs. 850°C-950°C), heat from exothermic carbonation cannot be directly used to provide the heat needed for endothermic calcination in the calciner. As a result, fuel (powdered coal) can directly be burned with oxygen provided by an air separation unit (ASU) in the calciner (as a fluidized-bed oxy-fuel combustor) to provide the necessary heat for calcination.



Compared to the market-ready MEA absorption process, lower capital cost, energy penalty, CO₂ capture cost, and less complex integration with the CO₂ emission source plant can be achieved by the calcium looping process (Cormos and Simon, 2013; Zhao et al., 2013). This happens due to i) the high availability of low-cost, non-hazardous, limestone-derived CaO sorbent from natural sources (Zhao et al., 2013), ii) no need for a separate FGD unit (as CaO itself can act as the desulphurization agent) (Zhao et al., 2013), and iii) the potential of spent CaO to be used elsewhere (mostly in the cement industry) (Dean et al., 2011a; Dean et al., 2011b). As a result, this technology emerged as a near-term alternative to the current MEA-based postcombustion CCT (Zhao et al., 2013).

The main concerns about the calcium looping process are the rapid decay in the CO₂ carrying capacity of CaO-based sorbents over multiple calcination-carbonation cycles and an efficient process design with optimal heat integration. Regarding sorbent deactivation, the decrease in the reactivity of CaO can be attributed to the i) sintering of particles during high-temperature carbonation-calcination processes, ii) particle attrition in fluidized beds, and iii) poisoning by sulfation/sulfidation reactions with SO_x, resulting in an outer sulfated layer forming on the particles, which prevents CO2 uptake (Sun et al., 2007; Manovic and Anthony, 2009; Wang et al., 2010). However, it has been reported that the presence of steam under real operating conditions significantly increases CaO sorbent activity, making CaO activity loss less of a concern (Manovic and Anthony, 2010; Donat et al., 2012). Regarding the necessity

of heat integration, it should be mentioned that providing the required heat for calcination through direct fuel burning under oxy-fuel conditions (by oxygen rather than air) prevents CO₂ dilution with N₂ and facilitates its purification. However, this coincides with the large energy consumption needed for the ASU, which results in a ~6% energy penalty (comparable with ~5% for CO_2 compression) (Abanades et al., 2007). Fortunately, excess heat recovery and integration into new steam cycles are possible through the process configuration presented in Figure 8, which can keep the energy penalty at an economically satisfactory level (Zhao et al., 2013). It has been reported that an 82% reduction in energy penalty and a 50% decrease in the cost of electricity increase of calcium looping can be achieved by integrating the potential heat that can be recovered from a calcium looping process used for postcombustion CO₂ capture from a power plant (Yang et al., 2010).

An alternative calcium looping process has been suggested (Junk et al., 2016) that could eliminate the need for an ASU; the combustion of additional fuel in a second combustor provides indirect heat for the calciner. The additional flue gas/CO₂ is combined with that of the main power plant. The cost of the heat transfer equipment is offset by the elimination of the ASU. The indirect heating methodology has been tested at a 300 kWth scale with plans for 20 MWth pilot plant (Hilz et al., 2019; Ströhle et al., 2021).

In this study, Tang et al.'s third case (Tang and You, 2018) is considered the representative Alt Tech for calcium looping technology.

	BAT	Alt Tech 1	Alt Tech 2	Alt Tech 3	
	MEA absorption	Membrane	Solid adsorption	Calcium looping	
TRL	9	7	4–5	6–7	
Pro	Commercially available	Cost-effective	Reduced energy penalty in regeneration	Any industrial CO ₂ sources Lower capital cost Lower energy penalty	
			Wider range of operating temp	Lower CO_2 capture cost Simple integration CO_2 emission plant	
			Less waste during cycling	FGD unit may not be needed	
		Near commercial production	Easier disposal of spent materials	Spent CaO as commodity for cement industry	
	Cost-effective	Economies of scope, similar technology to commercial RO	Low/no sorbent attrition in fixed beds		
Con	Energy cost to regenerate MEA	CO ₂ N ₂ selectivity	Solid adsorbents not yet cost- effective	Rapid decay in CO ₂ carrying capacity of CaO	
	MEA corrosive to plant			Need heat integration to drive Ca reactions	
	High level of desulfurization required High cost of captured CO ₂	Still relatively high power consumption	Sorbent attrition in recirculating beds	Huge volumes waste product if industrial sink not nearby	

TABLE 3 | Summary of BAT and three post-combustion CCTs as user's Alt Techs, TRL ranges, and pros and cons for those CCTs.

3.4 Qualitative Comparison Results (Input, Output, and Scoring)

Data entry for each CCT pathway is a key step in TDM development and is summarized in §3.4.1. Next, the key characteristics of each pathway are obtained, organized, and entered (§3.4.2). Finally, a radar plot comparing each Alt Tech to the BAT is generated (§3.4.3). The TDM tool and database were initially developed using Excel (§2.2, §2.3).

3.4.1 TDM Implementation-Data Input/Output

User data entry is in two areas, as shown in **Figure 2**. In the first steps (the first three columns on the left side of **Figure 2**), the user should select among different available options from the dropdown menus for the project scale, CO_2 source, and CO_2 target application, whereupon the TDM tool retrieves the corresponding KIP data values for the BAT and Target. The second data entry step is related to the KIP data values for the user's CCT Alt Tech to be studied (column 5 of **Figure 2**) and the year of the costing. Once there is enough data for all KIPs in the tables, relative comparisons can be made for the Alt Tech or user's CO_2 capture under study to the BAT and Target. Output results will be generated for further analysis to see where R&D resources could be reallocated. Samples of the qualitative outputs are presented in §3.4. As per §2.2.1, an inflation index is applied to CCCL TCR in generating the radar plots for the BAT and Alt Techs.

3.4.2 BAT and Alt Tech Database Generation

Key characteristics of the BAT and three Alt Techs described in the previous paragraphs are summarized in **Table 3**. This comparison highlights the importance of ongoing development to address current technology gaps. For each Alt Tech, a TRL range has been identified because it is quite difficult to precisely assign specific TRLs for non-commercial technologies. There are several criteria when identifying TRL (Buchner et al., 2018; Kearns, 2021).

These attributes need to be translated into measurable KIPs so they can be qualitatively and quantitatively analyzed by the TDM to support RD&D resource allocation decisions. The KIPs for the BAT and three Alt Techs are shown in **Table 4** and represent a subset of the KIPs extracted from available studies on these technologies. KIP values that had to be recalculated from TEA data in the publications referenced are indicated by underlined text. In addition to §2.2.2, this was to align KIPs and their units of measurement across CCTs studied in the TDM.

3.4.3 BAT and Alt Tech Comparison

By combining data from lab and pilot scales with those from simulations and TEA studies and then recalculating any missing KIP values from existing data, there is sufficient data in the literature for the BAT (Ciferno, 2007) and three Alt Techs (Sjostrom et al., 2016; Baker et al., 2018; Tang and You, 2018). However, knowledge and data gaps for the end application Target prevent a complete comparison of the Alt Techs against the BAT and Target. In terms of knowledge gaps, a few, if any, KIPs have been defined for performance, cost, or lifetime and then prioritized for a given application or desired market. In terms of data gaps, only a handful of Target values were found for those KIPs. One comparison can be made for the DOE target of no more than a 20% LCOE Increase with CC (US DOE/NETL, 2015b). In descending order and relative to the 20% LCOE Increase with the CC target: the BAT is 108% or a little more than five times; solid adsorption is 99% or a little less than five times; membrane is 78% or a little less than four times; and calcium looping is 42% or about double. Although the relative % LCOE Increase with CC is the least for calcium looping, the BAT and three Alt Techs all exceed the DOE target (US DOE/NETL, 2015b). The other Target, 90% CO₂ Recovery Efficiency, was only achieved for the BAT and solid adsorption. For membrane and calcium looping, their CCT performed best at CO₂ Recovery Efficiencies, less than 90%. A detailed explanation is given later in this section.

Any KIPs can be selected from Table 4, with at least three needed to generate a Radar Plot. Among the three equations

	Units	BAT	Alt Tech 1	Alt Tech 2	Alt Tech 3	Target
		MEA Absorption ^a	Membrane ^b	Solid Adsorption ^c	Calcium looping ^d	DOE
KIPs Performance Category						
Scale of CC Process	tCO ₂ /day	8,507	8,832	12,715	13,114	_
CC Energy Efficiency	kWh/tCO2_cap	368	328	<u>152</u>	275	_
CO ₂ Recovery Efficiency	%CO ₂ Rec	90	80	90	83	90
Net Energy/tCO ₂ Emitted before CC	kWh _{enet} _wo CC/ tCO ₂ _wo CC	1,101	1,221	1,165	1,047	_
Net Energy/tCO2 Emitted after CC	kWh _{enet} _w CC/tCO ₂ _w CC	7,702	4,792	9,343	4,850	_
HHV Efficiency before CC	%	35.0	35.0 ^f	36.8	37.2	-
HHV Efficiency after CC	%	24.5	27.5	29.5	29.8	-
% Loss Efficiency	%	30.0	21.5	19.9	19.9	_
LCOE before CC	c/kWh	6.40 ^g	5.00 ^h	7.21	5.27	_
LCOE Increase with CC	c/kWh	6.92	3.92	7.15	2.23	_
% LCOE Increase with CC	%	108	78	99	<u>42</u> 22.7	20
CO ₂ Capture Cost	\$/tCO ₂ Capt.	59.0	46.9	74.2	22.7	_
CO ₂ Avoidance or Mitigation Cost	\$/tCO ₂ Avoid.	89.0	64.1	95.2	30.5	_
Cost Category						
CCCL TCR ⁱ	\$/tCO ₂ /day	53,129	42,230	36,447	31,970	_
Variable O&M	\$/tCO ₂	31.58	22.80	29.85	17.16	_
Fixed O&M	\$/tCO ₂	27.64	24.10	44.22	5.52	_
Tech Replacement Costs	\$/tCO ₂	3.88	2.09	1.19	8.16	_
Lifetime Category						
Maximum Plant Life	Years	-	25	30	30	_
Levelization Period	Years	20	25	35	30	_
Key Tech Lifetime	Years	-	3.00	2.76	0.0057	_

TABLE 4 | TDM KIPs for the selected MEA scrubbing (BAT) and three post-combustion CCTs as user's Alt Techs of this work.

Underlined text indicates values that the authors recalculated from TEA data in publications referenced.

^aData from Ciferno (2007).

^bData from Baker et al. (2018).

^cData from Sjostrom et al. (2016).

^dData from Tang and You (2018).

^eData from US DOE/NETL, 2015b).

^fAssumed to be the same as BAT.

^gNew build on site.

^hElectricity purchased from grid.

ⁱInflation adjusted from the year of cost estimate in publication to 2020.

shown in §2.2.5, **Eq. 1** has been used. The BAT is normalized to serve as a comparison point in the radar plot. **Table 5** shows relative comparisons of each Alt Tech to the BAT using **Eq. 1**. In **Table 5**, six KIPs are chosen: CCUS Energy Efficiency, Net Energy/tCO₂ Emitted after CC, LCOE Increase with CC, CO₂ Avoidance/Mitigation Cost, CCCL TCR, and Tech Replacement Costs.

Comparisons can now be made among the KIPs for BAT, and three Alt Techs from **Table 4** are shown in **Table 5** and the radar plot in **Figure 9**.

Before progressing further in comparing the three Alt Techs based on **Table 5** and **Figure 9**, it is worth noting that the difference in TRLs could affect the relative quality of data inputs. Indeed, although the membrane technology is tested at a pre-commercial stage, this is not the case with calcium looping, which is often based on simulation and pilot scale data. In **Figure 9**, the KIP CCCL TCR is similar to all the CCTs studied. However, the other five KIPs showed differences in CCT performance and cost relative to the BAT. For CC Energy Efficiency, all Alt Techs were lower or more efficient than MEA, and membrane technology used 89% of the energy per tonne of CO₂ captured. Calcium looping used 75% compared to MEA, and solid adsorption (based on benzyl amide's ideal performance) consumed only 41% of the energy of MEA. All the Alt Techs had similar net energy production per tonne of CO2 emitted before CC, from approximately 1,100-1,220 kWhenet/tCO2 emitted. For Net Energy/tCO₂ Emitted after CC, solid adsorption is best, generating the most saleable power for a given amount of CO₂ emitted, approximately 20% more than MEA. Membrane technology and calcium looping are similar at 60% of MEAs performance. Note that both MEA and solid adsorption use the DOE Target of 90% CO2 Recovery Efficiency, whereas membrane and calcium looping do not. The "poor" performance of both membrane technology and calcium looping with this KIP is directly linked to the lower CO₂ capture rates of 80% and 83%, respectively. This halves the numeric value of this KIP because the energy production is normalized by the CO₂ emitted, not captured, or 20% rather than 10% for MEA and solid adsorption. However, membrane technology's CC Energy Efficiency, the first KIP, was minimized at 80% capture, not 90%. Junk et al. and Abanades et al. used

		Units	BAT/BAT	Alt Tech 1/BAT	Alt Tech 2/BAT	Alt Tech 3/BAT
KIPs	CC Energy Efficiency	%	100	89	41	75
	Net Energy/tCO2 Emitted after CC	%	100	62	121	63
	LCOE Increase with CC	%	100	57	103	32
	CO ₂ Avoidance or Mitigation Cost	%	100	72	107	34
	CCCL TCR	%	100	79	69	60
	Tech Replacement Costs	%	100	54	31	210

TABLE 5 | Relative comparisons of three Alt Techs of this study to BAT using Eq. 1 for the selected KIPs from Table 4.

BAT is MEA absorption, Alt Tech 1 is membrane, Alt Tech 2 is solid adsorption, and Alt Tech 3 is calcium looping.



 CO_2 capture rates of 80% and 86%, respectively (Abanades et al., 2007; Junk et al., 2016). Therefore, the calcium looping penalty on this KIP is not unique to Tang and You's study (Tang and You, 2018).

Regarding LCOE Increase with CC, calcium looping was the lowest at 32% of BAT, then membrane technology with a 57% increase, and finally solid adsorption at 103%. When normalized by the % LCOE Increase for MEA as the BAT (108%), the relative performance between the Alt Techs is, of course, unchanged and is similar because the % LCOE Increase of the BAT was close to 100%.

The same trend can be seen for CO_2 Avoidance or Mitigation Cost. Calcium looping is dramatically lower at 34% than the BAT. The membrane is also lower than the BAT at 72% but more than double that for calcium looping. Solid adsorption is slightly more than the BAT at 107%. The similarity between the LCOE Increase with CC and CO_2 Avoidance or Mitigation Cost is not surprising because the LCOE is a component in the equation to calculate CO_2 Avoidance or Mitigation Cost.

CCCL TCR based on the CC rate of 1 tCO_2/day was the lowest for calcium looping at 60%, followed by solid

adsorption at 69% of the BAT, respectively. The membrane was the highest at 79% of the BAT. The BAT and solid adsorption studies listed the Bare Erected Cost (BEC), Total Installed Cost (TIC), and, finally, the highest TCR. The membrane and calcium looping studies listed total capital or TIC, and so it is assumed for the purposes of comparisons in this work that they also used the same cost basis because these costs were used to determine their capture or avoided CO_2 costs. However, if the membrane and calcium studies did not include contingencies and overhead costs, the TCR for these may be underestimated.

The greatest relative difference is for Tech Replacement Costs, where calcium looping is more than double, or the most expensive, relative to MEA, whereas the other two Alt Techs have lower costs relative to MEA. The membrane has high initial costs, but its technology lifetime was specified at 3 years by MTR (Baker et al., 2018). The solid sorbent cost is very high, \$US 10,100/tonne, compared to limestone, \$US 25/tonne, but the makeup of limestone per tonne of CO_2 is much higher due to CaO deactivation.

The next step is to confirm what is causing the relative differences seen in the radar plot. Particularly for new

TABLE 6 Comparison of calcium looping KIPs as a function of various levelization
periods, interest rates and VOM, FOM charge factors.

	Units	Calcium looping financial assumptions			
		Tang	BAT	BAT, Tang	
Levelization Period	years	30 ^a	20 ^b	20 ^b	
Interest Rate	%	7.0 ^a	16.7 ^b	16.7 ^b	
Fixed Charge Factor	n/a	0.0806 ^a	0.1750 ^b	0.1750 ^b	
VOM & FOM Charge Factor	n/a	1.000 ^a	1.157 ^b	1.000 ^a	
CO ₂ Avoidance Cost ^a	\$/tCO2	29.8	72.2	36.7	
CO ₂ Capture Cost ^a	\$/tCO2	22.7	54.8	27.9	
LCOE Before CC ^a	c/kWh	5.265	8.789	7.919	
LCOE Increase with CC ^a	c/kWh	2.234	2.631	2.749	
% LCOE Increase with CC ^a	%	42.4	36.1	34.7	

Column "Tang" uses Tang's published values. Column "BAT" uses Tang's capital, VOM, and FOM, but with the BAT's interest rates, amortization period, and VOM/FOM charge factors. Column "BAT, Tang" uses the BAT interest and amortization period and no VOM/ FOM charge factors as per Tang.

^aData from Tang and You (2018).

^bData from Ciferno (2007). Bold values indicate KIPs from **Table 5** and **Figure 9** that change depending on different financial assumptions used for FCF and VOM & FOM Charge Factor.

research fields and low TRLs, one must ask if TEA source data, calculations, equations, and assumptions are valid and equal. It follows that TDM qualitative outputs are another tool to check for knowledge and data gaps, or how the TEA was done and how it compares to those for similar technologies. This enables standardization or consensus among CCUS researchers and serves as a way to make more accurate, transparent comparisons of knowledge, data, and technology gaps and hence another way to help accelerate CCUS RD&D. Future work could explore validating TEA source data used in **Figure 9** to see if and how differences in TEA assumptions among CCT affect results used in the TDM.

For example, potential variations in the TCR (or overnight costs, investment costs, etc.) were discussed above. However, it is not always clear if all contingencies are included or what factors are used for those contingencies. The BAT contained detailed costs for, using that studies terminology, major Capital Costs, BEC, and Total Investment Cost (not TIC). The BEC was 212%, and the Total Investment Cost was 370% of the major Capital Costs, based on rates for that study. The process and project contingencies were 25% and 18% of the BEC in that study. Different values may be used in different TEAs and for different technologies, and Total Investment Cost and TCR are interchangeably used in different studies.

Similar questions regarding comparisons with different assumptions were raised when the KIPs for CO_2 Avoidance or Mitigation Cost and LCOE Increase with CC were calculated. The TCR will impact the LCOE in terms of not only the capital value but also the Fixed Charge Factor (FCF). The FCF is a multiplier for the TCR, accounting for the levelization or financing period and the interest rate assumed.

The impact of these different financial assumptions is summarized in **Table 6**. The Tang calcium looping TEA used a levelization period of 30 years and an interest rate of 7.0%, or an FCF of 0.0806 (Tang and You, 2018). The BAT baseline reported an FCF of 0.1750, corresponding to 15 years and 15.5% interest (Ciferno, 2007). The BAT also applied levelization charge factors of 1.157 to their Variable Operation and Maintenance (VOM) and Fixed Operation and Maintenance (FOM). When those BAT financing assumptions are used for the calcium looping case, the CO₂ Avoidance or Mitigation Cost increased from 29.8 to 72.2 \$/tCO2. Changes to other KIPs are in column "BAT." In order to compare the effect of different financial assumptions on CO₂ Avoidance or Mitigation Costs in Table 6, values were calculated using Tang et al.'s LCOE equations and CO₂ produced. Therefore, the CO₂ Avoidance or Mitigation Cost in Table 6 is different from that in Table 4. The calculated LCOE values, however, matched the reported values to within 1 c/kWh.

The % LCOE Increase with CC did not change dramatically because the BAT financial parameters were also applied to the new reference power plant without CC. Furthermore, the impact of the VOM and FOM Charge Factor is greater than that of plant financing. Assuming the BAT FCF and Tang's value of 1 for VOM and FOM Charge Factor, the CO₂ Avoidance Cost increased from 29.8 to 36.7 \$/tCO₂, column "BAT, Tang."

In the previous paragraphs, we used several KIPs to compare different CCTs and generated Tables 4-6 for this. We need an additional step to capture and integrate all the outputs to facilitate decision-making. The whole TDM process of determining KIPs, units of measurement, measurement standards, tabulating KIP data, and then making qualitative comparisons in radar plots can be useful for RD&D at each TRL stage. It can be a facile process for users to check data from TEA and or measured data to uncover knowledge gaps at lower TRLs (1-3) through measurement and data gaps at mid-TRLs (4-6) and, finally, technology gaps at higher TRLs (7-9). Thus, at each stage of RD&D, researchers can use the TDM to make valid comparisons, support decision-making, and accelerate commercial deployment of the most promising technologies for a given end application.

The current TDM assumes a relative weighting of 1.0 for each KIP. Future TDM versions could include adding a range of KIP weightings to correspond to prioritizing KIPs for a given CCT application. Some decision-making tools include KIP weightings of 0.0–2.0 in increments of 0.5, and something similar could be done for the CCT TDM (ES-SelectTM, 2012). Another approach is to 1) identify the minimal requirement for each KIP and 2) use cost per unit CO₂ capture (additional cost of per unit electricity production) as the final ranking. A recent IEA report showed this approach (Baylin-Stern and Berghout, 2021).

Of interest, the US DOE has updated its CCUS target using a new indicator based on the cost of per unit CO_2 mitigation (DOE/EE-2362, 2021). Often referred to as Levelized Cost of Carbon Abatement (LCCA) (Friedmann et al., 2020), this provides a good indicator to compare different technologies, including CCUS. This indicator is more suitable for assessing

policy incentives for CO_2 mitigation. However, this indicator could not be used in this study as CO_2 capture needs to be combined with CO_2 storage and utilization stages. This work is ongoing at NRC to integrate utilization and storage technology options in the proposed TDM. As shown in **Figure 2**, the proposed TDM framework is designed to compare different CCUS pathways.

All four CCTs discussed in this study have progressed past the lab scale (TRL 1-3), so their fundamental technologies are essentially unchanged from the pilot scale (TRL 4-6) onward. Thus, there are limited opportunities for disruptive or stepchange improvements. However, commercial deployment and resulting economies of scale could still lead to per-unit cost reductions while decreasing GHG emissions. System optimization could lead to marginal reductions in cost and GHG emissions. Materials' cost alone will have some impact on the CO₂ Capture Cost. Furthermore, future improvements in both cost and performance of functional materials could have a significant, step-change impact when incorporated into existing high TRL systems.

As an example, the impact of the key technology's materials cost is discussed first, and its performance is second. First, MTR membrane materials account for $6.4 \$ /tCO₂ captured or 13.6% of the total 47 $\$ /tCO₂ captured. If the membrane cost were halved to $3.2 \$ /tCO₂ captured, then the CO₂ capture cost would decrease by at least 6.8%. However, the 47 $\$ /tCO₂ capture cost also includes 19 $\$ /tCO₂ captured of interest or depreciation. Because membranes represent 31% of initial capital cost and the total capital expenses, a 50% reduction of membrane costs also corresponds to 50% of 31% of 19 $\$ /tCO₂ captured or 2.95 $\$ /tCO₂ captured representing an overall reduction of $3.2 + 3.0 = 6.2 \$ /tCO₂ captured, which is a more significant contribution.

Second, additional improvements could be achieved through increased materials performance. Again, using the MTR membrane includes a higher capture rate (permeance and/or specificity), longer lifetime, and/or lower temperature. These parameters could reduce energy input requirements. MTR has already developed second- and third-generation membranes, which have double and triple the throughput with the same selectivity. Doubling the throughput is analogous to the discussion above, where the specific membrane cost is halved but also has an additional impact on capital costs with fewer pressure vessels and the associated installation infrastructure, as well as piping.

Ciferno's study, used as the BAT, compared the cost of three generations of amines with different regeneration energy requirements and their impact on the PC power plant's net power output (Ciferno, 2007). Case 1 in the 2007 study is used in this study, and Case 1a is the same plant but with lower regeneration energy. The new amine reduced the LCOE incremental cost from 69.2 to 63.2 %/MWh, and the CO₂ avoided cost decreased from 89 to 81 \$/tCO₂ (currencies in US dollars from the year of the study) (Ciferno, 2007). In summary, adding up these material costs and performance improvements could result in significant improvement in CCT's overall cost and GHG emissions. This indicates the relative importance of improving

functional materials cost and performance in future commercial CCT deployment.

4 DISCUSSION

The concept, framework, and methodology were established, and a new decision-making tool, a TDM, was developed using Excel and then applied to CCUS and specifically to CCTs. The TDM is a technology screening and data inventory tool to perform complimentary analysis and assessment with TEA and LCA tools to evaluate the cost, GHG, and market impacts of both disruptive CCUS technologies and their novel materials. The TDM can also be used to cross-check several CCTs and support continuous improvement of the respective TEA methods, calculations, and data.

KIPs across three categories of performance, cost, and lifetime at the same TRL, representing the entire CCT process being studied, were introduced to measure and compare among CCTs consisting of the BAT, Alt Tech under study, and the end application Target or Goal it will be used for. A PC-fired power plant retrofitted with commercially available monoethanolamine (MEA) CO_2 capture absorption technology served as the BAT. Three promising CCT Alt Techs were presented: membrane, solid adsorption, and calcium looping.

Collecting KIP data for the BAT, Alt Techs, and Target revealed knowledge (how to measure a technology) and data (corresponding data measurement) gaps. Although there are international organizations active in this field, a knowledge gap is at this relatively early stage of CO_2 capture research. There is not yet an agreed set of KIPs, and neither the units of measurement nor the measurement standards have been finalized. The corresponding data gaps are limited operational or *in situ* data at the lab bench top, pilot, or demo scales for both promising CCTs Alt Techs under development and CCT Targets. Specifically, a lack of a complete data set for CCT end application Targets or goals prevented TDM comparisons among BAT, Alt Tech, and Target.

The TDM's KIP data table and resulting qualitative radar chart were demonstrated using available data for the BAT and three Alt Techs. This was accomplished by combining data from lab and pilot scales with those from simulations and TEA studies and then recalculating any missing KIP values from existing data. Relative comparisons were then made by normalizing each Alt Tech to the BAT for six KIPs to generate the radar plot. CCT performance and cost were compared among the four CCTs, and differences were easily visualized. However, these apparent differences must then be cross-checked against the original TEA data to understand if they are due to the CCT itself (technology gaps) or from the experimental or operational measurements (data gaps), or assumptions, equations, and KIPs used to generate the TEA (knowledge gaps). An examination of TEA financial assumptions revealed knowledge gaps of how different levelization periods, interest rates, FOM, and VOM charge factors impact CO₂ Avoidance Cost and LCOE increase. Future work includes further TDM development, including

using software other than Excel, integrating the TDM with NRC's webserver, validating across TEA source data, TDM comparisons within the same CCT, improving KIPs, introducing KIP weighting, and introducing quantitative TDM outputs, including technology ranking based on a combined scoring factor, obtaining end application Target or Goal data, and finally expanding the TDM to include carbon conversion and utilization.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/**Supplementary Material**. Further inquiries can be directed to the corresponding author.

AUTHOR CONTRIBUTIONS

RB, OA, MMD-C, and FB were all involved in the writing and editing of the article. All authors approved the final version of the article.

REFERENCES

- Abanades, J. C., Grasa, G., Alonso, M., Rodriguez, N., Anthony, E. J., and Romeo, L. M. (2007). Cost Structure of a Postcombustion CO2 Capture System Using CaO. *Environ. Sci. Technol.* 41, 5523–5527. doi:10.1021/es070099a
- Alstom UK (2011). European Best Practice Guidelines for Assessment of CO₂ Capture Technologies. Trondheim Nor.SINTEF 4.3, 1–112. doi:10.5281/ zenodo.1312801
- Anantharaman, R., Berstad, D., and Roussanaly, S.The International Conference on Applied Energy, I. (2014). Techno-economic performance of a hybrid membrane - Liquefaction process for post-combustion CO₂ capture. *Energy Procedia* 61, 1,244–1,247.doi:10.1016/j.egypro.2014.11.1068
- Anantharaman, R., Peters, T., Xing, W., Fontaine, M.-L., and Bredesen, R. (2016). Dual Phase High-Temperature Membranes for CO2 Separation - Performance Assessment in Post- and Pre-combustion Processes. *Faraday Discuss.* 192, 251–269. doi:10.1039/c6fd00038j
- Baker, R. W., Freeman, B., Kniep, J., Huang, Y. I., and Merkel, T. C. (2018). CO2 Capture from Cement Plants and Steel Mills Using Membranes. *Ind. Eng. Chem. Res.* 57, 15963–15970. doi:10.1021/acs.iecr.8b02574
- Baylin-Stern, A., and Berghout, N. (2021). Is Carbon Capture Too Expensive? [Online]. International Energy Agency. Paris, France, Available: https://www. iea.org/commentaries/is-carbon-capture-too-expensive [Accessed 20 January 2022].
- Biegler, L. T., and Grossmann, I. E. (2004). Retrospective on Optimization. Comput. Chem. Eng. 28, 1169–1192. doi:10.1016/j.compchemeng.2003.11.003
- Board, O. S.National Academies of Sciences, E., and Medicine (2019). Negative Emissions Technologies and Reliable Sequestration: A Research Agenda. National Academies Press. Washington, D.C., United States.
- Buchner, G. A., Zimmermann, A. W., Hohgräve, A. E., and Schomäcker, R. (2018). Techno-economic Assessment Framework for the Chemical Industry-Based on Technology Readiness Levels. *Ind. Eng. Chem. Res.* 57, 8502–8517. doi:10.1021/ acs.iecr.8b01248
- Chen, Z. (2022). A Review of Pre-combustion Carbon Capture Technology, Proceedings of the 2022 7th International Conference on Social Sciences and Economic Development (ICSSED 2022), Wuhan, China, March 25–27, 2022. Atlantis Press, 524–528.
- Ciferno, J. P. (2007). Carbon Dioxide Capture from Existing Coal-Fired Power Plants. National Energy Technology Laboratory. Pittsburgh Pennsylvania, United States.

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SUPPLEMENTARY MATERIAL

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- Cormos, A., and Simon, A. (2013). Dynamic Modelling of CO₂ Capture by Calcium-Looping Cycle. *Chem. Eng. Trans.* 35, 421–426. doi:10.3303/ CET1335070
- Dean, C. C., Blamey, J., Florin, N. H., Al-Jeboori, M. J., and Fennell, P. S. (2011a). The Calcium Looping Cycle for CO2 Capture from Power Generation, Cement Manufacture and Hydrogen Production. *Chem. Eng. Res. Des.* 89, 836–855. doi:10.1016/j.cherd.2010.10.013
- Dean, C. C., Dugwell, D., and Fennell, P. S. (2011b). Investigation into Potential Synergy between Power Generation, Cement Manufacture and CO2 Abatement Using the Calcium Looping Cycle. *Energy Environ. Sci.* 4, 2050–2053. doi:10. 1039/c1ee01282g
- Dillon, D., Swisher, J., Freeman, B., Mckaskle, R., Chu, R., Lu, H., et al. (2021).Engineering Design of a Full Scale Membrane Post-Combustion CO₂ Capture System Retrofit for Duke Energy's East Bend Station, 15th International Conference on Greenhouse Gas Control Technologies, GHGT-15, Abu Dhabi, UAE, March 15–18, 2021. SSRN.
- DOE/EE-2362 Carbon Negative Shot [Online] (2021). United States: US Department of Energy. Available: https://www.energy.gov/fecm/carbon-negative-shot [Accessed 20 November 2021].
- Donat, F., Florin, N. H., Anthony, E. J., and Fennell, P. S. (2012). Influence of High-Temperature Steam on the Reactivity of CaO Sorbent for CO2 Capture. *Environ. Sci. Technol.* 46, 1262–1269. doi:10.1021/es202679w
- ES-Select[™] (2012). Documentation and User's Manual, Ver. No. 2.0. Department of Energy, Sandia National Lab. Albuquerque, New Mexico, United States.
- Freeman, B., Hao, P., Baker, R., Kniep, J., Chen, E., Ding, J., et al. (2014). Hybrid Membrane-Absorption CO2 Capture Process. *Energy Procedia* 63, 605–613. doi:10.1016/j.egypro.2014.11.065
- Friedmann, J., Fan, Z., Byrum, Z., Ochu, E., Bhardwaj, A., and Sheerazi, H. (2020). Levelized Cost of Carbon Abatement: An Improved Cost-Assessment Methodology for a Net-Zero Emissions World. New York, NY, USA: Columbia University SIPA Center on Global Energy Policy.
- Haszeldine, R. S., Flude, S., Johnson, G., and Scott, V. (2018). Negative Emissions Technologies and Carbon Capture and Storage to Achieve the Paris Agreement Commitments. *Phil. Trans. R. Soc. A* 376, 20160447. doi:10.1098/rsta.2016.0447
- Hilz, J., Haaf, M., Helbig, M., Lindqvist, N., Ströhle, J., and Epple, B. (2019). Scaleup of the Carbonate Looping Process to a 20 MWth Pilot Plant Based on Long-Term Pilot Tests. *Int. J. Greenh. Gas Control* 88, 332–341. doi:10.1016/j.ijggc. 2019.04.026
- Hochgesand, G. (1970). Rectisol and Purisol. Ind. Eng. Chem. 62, 37–43. doi:10. 1021/ie50727a007

- Hovington, P., Ghaffari-Nik, O., Mariac, L., Liu, A., Henkel, B., and Marx, S. (2021). Rapid Cycle Temperature Swing Adsorption Process Using Solid Structured Sorbent for CO₂ Capture from Cement Flue Gas. Proceedings of the 15th Greenhouse Gas Control Technologies Conference, 15-18 March 2021, *Available at SSRN 3814414*. Available at: http://dx.doi.org/10.2139/ssrn. 3814414 (Accessed July 12, 2022).
- Huang, Y., Rebennack, S., and Zheng, Q. P. (2013). Techno-economic Analysis and Optimization Models for Carbon Capture and Storage: a Survey. *Energy Syst.* 4, 315–353. doi:10.1007/s12667-013-0086-0
- Junk, M., Reitz, M., Ströhle, J., and Epple, B. (2016). Technical and Economical Assessment of the Indirectly Heated Carbonate Looping Process. J. Energy Resour. Technol. 138. 4033142. doi:10.1115/1.4033142
- Kearns, D., Liu, H., and Consoli, C. (2021). TECHNOLOGY READINESS AND COSTS OF CCS. Melbourne Australia: Global CCS Institute.
- Kniep, J., Baker, R., Casillas, C., Chan, K., Huang, I., Merkel, T., et al. (2018). "Small Pilot Test Results from a Polaris Membrane 1 MWe CO₂ Capture System," in 14th Greenhouse Gas Control Technologies Conference Melbourne, Melbourne, Australia, October 21–25, 2018, 21–26.
- Krutka, H., Sjostrom, S., Starns, T., Dillon, M., and Silverman, R. (2013). Postcombustion CO2 Capture Using Solid Sorbents: 1 MWe Pilot Evaluation. *Energy Procedia* 37, 73–88. doi:10.1016/j.egypro.2013.05.087
- Kuramochi, T., Ramírez, A., Turkenburg, W., and Faaij, A. (2012). Comparative Assessment of CO2 Capture Technologies for Carbon-Intensive Industrial Processes. *Prog. Energy Combust. Sci.* 38, 87–112. doi:10.1016/j.pecs.2011. 05.001
- Leung, D. Y. C., Caramanna, G., and Maroto-Valer, M. M. (2014). An Overview of Current Status of Carbon Dioxide Capture and Storage Technologies. *Renew.* Sustain. Energy Rev. 39, 426–443. doi:10.1016/j.rser.2014.07.093
- Lin, J.-B., Nguyen, T. T. T., Vaidhyanathan, R., Burner, J., Taylor, J. M., Durekova, H., et al. (2021). A Scalable Metal-Organic Framework as a Durable Physisorbent for Carbon Dioxide Capture. *Science* 374, 1464–1469. doi:10. 1126/science.abi7281
- Malek, K., and Nathwani, J. (2015). "Technology Management Tools for Assessing Emerging Technologies: The Case of Grid-Scale Storage", in: 2015 Portland International Conference on Management of Engineering and Technology (PICMET), 02-06 Aug. 2015, Portland, OR, USA, 2346–2354.
- Malek, K. (2019). Typology of Business Models for Emerging Grid-Scale Energy Storage Technologies. Waterloo, Ontario: UWSpace. Available at: http://hdl. handle.net/10012/14503
- Manovic, V., and Anthony, E. J. (2010). Carbonation of CaO-Based Sorbents Enhanced by Steam Addition. *Ind. Eng. Chem. Res.* 49, 9105–9110. doi:10.1021/ ie101352s
- Manovic, V., and Anthony, E. J. (2009). Long-Term Behavior of CaO-Based Pellets Supported by Calcium Aluminate Cements in a Long Series of CO2 Capture Cycles. *Ind. Eng. Chem. Res.* 48, 8906–8912. doi:10.1021/ie9011529
- McNeil, L. A., Mutch, G. A., Iacoviello, F., Bailey, J. J., Triantafyllou, G., Neagu, D., et al. (2020). Dendritic Silver Self-Assembly in Molten-Carbonate Membranes for Efficient Carbon Dioxide Capture. *Energy Environ. Sci.* 13, 1766–1775. doi:10.1039/c9ee03497h
- Merkel, T., Amo, K., Baker, R., Daniels, R., Friat, B., He, Z., et al. (2009). *Membrane Process to Sequester CO₂ from Power Plant Flue Gas.* USA: Membrane Technology and Research Inc.
- Merkel, T. C., Lin, H., Wei, X., and Baker, R. (2010). Power Plant Post-combustion Carbon Dioxide Capture: An Opportunity for Membranes. J. Membr. Sci. 359, 126–139. doi:10.1016/j.memsci.2009.10.041
- Merkel, T., Kniep, J., and Hofmann, T. (2018). "Scale up and Testing of Advanced Polaris Membrane CO2 Capture Technology (DE FE0031591),". Editor U. S. D. O. E. U. Doe (Pittsburgh PA: DOE National Energy Technology Laboratory (NETL)).
- NRC (2022). Economic & Environmental Evaluation of Carbon Conversions [Online]. Government of Canada. Available: https://eeecc.nrc-cnrc.gc.ca [Accessed June 14 2022].
- Pardakhti, M., Jafari, T., Tobin, Z., Dutta, B., Moharreri, E., Shemshaki, N. S., et al. (2019). Trends in Solid Adsorbent Materials Development for CO2 Capture. ACS Appl. Mat. Interfaces 11, 34533–34559. doi:10.1021/acsami.9b08487
- Qiao, Z., Zhao, S., Sheng, M., Wang, J., Wang, S., Wang, Z., et al. (2019). Metal-induced Ordered Microporous Polymers for Fabricating Large-Area Gas Separation Membranes. *Nat. Mater* 18, 163–168. doi:10.1038/s41563-018-0221-3

- Rezaei, F., and Webley, P. (2009). Optimum Structured Adsorbents for Gas Separation Processes. Chem. Eng. Sci. 64, 5182–5191. doi:10.1016/j.ces.2009.08.029
- Rubin, E., Booras, G., Davison, J., Ekstrom, C., Matuszewski, M., Mccoy, S., et al. (2013a). Toward a Common Method of Cost Estimation for CO₂ Capture and Storage at Fossil Fuel Power Plants. Australia: Global CCS Institute.
- Rubin, E. S., Short, C., Booras, G., Davison, J., Ekstrom, C., Matuszewski, M., et al. (2013b). A Proposed Methodology for CO2 Capture and Storage Cost Estimates. *Int. J. Greenh. Gas Control* 17, 488–503. doi:10.1016/j.ijggc.2013. 06.004
- Samanta, A., Zhao, A., Shimizu, G. K. H., Sarkar, P., and Gupta, R. (2012). Postcombustion CO2 Capture Using Solid Sorbents: A Review. *Ind. Eng. Chem. Res.* 51, 1438–1463. doi:10.1021/ie200686q
- Scholes, C. A., Anderson, C. J., Cuthbertson, R., Stevens, G. W., and Kentish, S. E. (2013). Simulations of Membrane Gas Separation: Chemical Solvent Absorption Hybrid Plants for Pre- and Post-Combustion Carbon Capture. *Sep. Sci. Technol.* 48, 1954–1962. doi:10.1080/01496395.2013.793201
- Shao, P., Dal-Cin, M. M., Guiver, M. D., and Kumar, A. (2013). Simulation of Membrane-Based CO2 Capture in a Coal-Fired Power Plant. J. Membr. Sci. 427, 451–459. doi:10.1016/j.memsci.2012.09.044
- Shi, S. Y., Jianjun Gonzales-Calienes, Giovanna, Baker, Ryan, Malek, Kourosh, and Skrivan, Will (2020). Canadian Energy Storage Report: 2019 Case Study for the Ontario Market. NRC Publications Archive: National Research Council of Canada. Ottawa, Ontario, Canada.
- Shimizu, T., Hirama, T., Hosoda, H., Kitano, K., Inagaki, M., and Tejima, K. (1999). A Twin Fluid-Bed Reactor for Removal of CO2 from Combustion Processes. *Chem. Eng. Res. Des.* 77, 62–68. doi:10.1205/026387699525882
- Sjostrom, S., Denney, J., Senior, C., and Morris, W. (2016). Evaluation of Solid Sorbents as a Retrofit Technology for CO₂ Capture. ADA-ES, Highlands Ranch, CO: US Department Of Energy Topical Report: US Department Of Energy Final Technical Report. DOE DE-FE004343 [Online]. United States. Available at: https://www.osti.gov/servlets/ purl/1261627 [Accessed 20 January 2022].
- Stevenson, M., and Armpriester, A. (2016). NRG CO2NCEPT Confirmation of Novel Cost-Effective Emerging Post-combustion Technology. Houston, Texas, United States: NRG Energy, Inc. and USDOE Office of Fossil Energy.
- Ströhle, J., Hofmann, C., Greco-Coppi, M., and Epple, B. (2021).CO₂ Capture from Lime and Cement Plants Using an Indirectly Heated Carbonate Looping Process–The Anica Project, TCCS–11. CO2 Capture, Transport and Storage. Short Papers from the 11th International Trondheim CCS Conference, Trondheim, 22nd–23rd June 2021. SINTEF Academic Press.
- Sun, P., Grace, J. R., Lim, C. J., and Anthony, E. J. (2007). The Effect of CaO Sintering on Cyclic CO2 Capture in Energy Systems. *AIChE J.* 53, 2432–2442. doi:10.1002/aic.11251
- Tang, Y., and You, F. (2018). Life Cycle Environmental and Economic Analysis of Pulverized Coal Oxy-Fuel Combustion Combining with Calcium Looping Process or Chemical Looping Air Separation. J. Clean. Prod. 181, 271–292. doi:10.1016/j.jclepro.2018.01.265
- Tapia, J. F. D., Lee, J.-Y., Ooi, R. E. H., Foo, D. C. Y., and Tan, R. R. (2018). A Review of Optimization and Decision-Making Models for the Planning ofCO2capture, Utilization and Storage (CCUS) Systems. Sustain. Prod. Consum. 13, 1–15. doi:10.1016/j.spc.2017.10.001
- Tian, Q., Zhao, D., Li, Z., Zhang, J., Li, Q., Liu, H., et al. (2016). Modelling and Evaluating CCUS: a Survey. *Ijcat* 53, 1–12. doi:10.1504/ijcat.2016. 073607
- US DOE/NETL (2015a). DOE/NETL Carbon Capture Program: Carbon Dioxide Capture Handbook [Online]., Available: https://netl.doe.gov/sites/default/files/ netl-file/Carbon-Dioxide-Capture-Handbook-2015.pdf [Accessed 15 March 2021].
- US DOE/NETL (2015b). Funding Opportunity Announcement DE-FOA-0001190 Entitled "Small And Large Scale Pilots For Reducing The Cost Of CO₂ Capture And Compression [Online]., Available at: https://www.netl.doe. gov/node/8590 [Accessed 20 January 2022].
- Wang, B., Qiao, Z., Xu, J., Wang, J., Liu, X., Zhao, S., et al. (2020). Unobstructed Ultrathin Gas Transport Channels in Composite Membranes by Interfacial Self-Assembly. Adv. Mat. 32, 1907701. doi:10. 1002/adma.201907701

- Wang, J., Manovic, V., Wu, Y., and Anthony, E. J. (2010). A Study on the Activity of CaO-Based Sorbents for Capturing CO2 in Clean Energy Processes. Appl. Energy 87, 1453–1458. doi:10.1016/j.apenergy.2009.08.010
- Wei, X., Baker, R. W., Merkel, T. C., and Freeman, B. C. (2015). Sweep-based Membrane Gas Separation Integrated with Gas-Fired Power Production and CO₂ Recovery. U.S. Patent No 9. Washington, DC: U.S. Patent and Trademark Office, 140–186.
- Yang, Y., Zhai, R., Duan, L., Kavosh, M., Patchigolla, K., and Oakey, J. (2010). Integration and Evaluation of a Power Plant with a CaO-Based CO2 Capture System. Int. J. Greenh. Gas Control 4, 603–612. doi:10.1016/j.ijggc.2010.01.004
- Zhao, M., Minett, A. I., and Harris, A. T. (2013). A Review of Techno-Economic Models for the Retrofitting of Conventional Pulverised-Coal Power Plants for Post-combustion Capture (PCC) of CO2. *Energy Environ. Sci.* 6, 25–40. doi:10. 1039/c2ee22890d
- Zimmermann, A., Müller, L. J., Marxen, A., Armstrong, K., Buchner, G., Wunderlich, J. K., et al. (2018). *Techno-Economic Assessment & Life-Cycle Assessment Guidelines for CO₂ Utilization*. Sheffield UK: CO2 Chem Media and Publising Ltd.
- Zimmermann, A., Müller, L., Wang, Y., Langhorst, T., Wunderlich, J., Marxen, A., et al. (2020a). Techno-Economic Assessment & Life Cycle Assessment Guidelines for CO₂ Utilization (Version 1.1). Ann Arbour, Michigan: Global CO2 Initiative at the University of Michigan and EIT Climate-KIC.

Zimmermann, A. W., Wunderlich, J., MüLler, L., Buchner, G. A., Marxen, A., Michailos, S., et al. (2020b). Techno-Economic Assessment Guidelines for CO₂ Utilization. *Front. Energy Res.* 8, 1–23. doi:10.3389/fenrg.2020.00005

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NOMENCLATURE

Alt Tech(s) Alternative Technology(ies) ASU Air Separation Unit BAT Best Available Technology BEC Bare Erected Costs **BoP** Balance of Plant BoS Balance of System **CAPEX** CAPital Expenditure CC Carbon Capture CCCL Carbon Capture Compression and Liquefaction CCUS Carbon capture, utilization, and storage **CCT(s)** Carbon Capture Technology(ies) CFB Circulating Fluidized Bed DAC Direct Air (carbon) Capture **DOE** Department of Energy EOR Enhanced Oil Recovery ES Energy Storage ESP Electro-Static Precipitator FCF Fixed Charge Factor FGD Flue Gas Desulfurization FOA Funding Opportunity Announcement FOM Fixed Operation and Maintenance **GCI** Global CO₂ Initiative GHG Green House Gas GPU Gas Permeation Units HHV Higher Heating Value IEA International Energy Agency IGCC Integrated Gas Combined Cycle

KIP(s) Key Indicator Parameter(s) LCA Life-cycle assessment LCCA Levelized Cost of Carbon Abatement LCOE Levelized Cost of Electricity LDH Layered Double Hydroxide LHV Lower Heating Value MEA MonoEthanol Amine MOF Metal–Organic Framework MTR Membrane Technology Research NETL National Energy Technology Laboratory NRC National Research Council of Canada **O&M** Operation and Maintenance **OPEX** Operational EXPenditure PC Pulverized Coal **PM** Particulate Matter PEMFC Proton Exchange Membrane Fuel Cell **R&D** Research and Development RD&D Research, Development and Deployment SCAH Steam Coil Air Heater TDM Technology Development Matrix TEA Techno-Economic Analysis TFC Thin Film Composite TCR Total Capital Required TIC Total Installed Cost **TRL(s)** Technology Readiness Level(s) TSA Temperature Swing Adsorption **US** United States VOM Variable Operation and Maintenance