



Identifying geologic characteristics and operational decisions to meet global carbon sequestration goals

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TITLE

Identifying geologic characteristics and operational decisions to meet global carbon sequestration goals

AUTHORS

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ABSTRACT

Geologic carbon sequestration is the process of injecting and storing CO₂ in subsurface reservoirs and is an essential technology for global environmental security (e.g., climate change mitigation) and economic security (e.g., CO₂ tax credits). To meet energy, economic, and environmental goals, society will have to identify vast volumes of high-capacity, low-cost, and viable storage reservoirs for sequestering CO₂. In turn, this requires understanding how major geologic characteristics (such as reservoir depth, thickness, permeability, porosity, and temperature) and design and operational decisions (such as injection well spacing) impact CO₂ injection rates, storage capacity, and economics. Although many numerical simulation tools exist, they cannot repeat the required thousands or millions of simulations to identify ideal reservoir properties and the sensitivity and interaction between geologic parameters and operational decisions. Here, we use *SCO₂T* (pronounced “Scott”; Sequestration of CO₂ Tool)—a fast-running, reduced-order modeling framework—to explore the sensitivity of major geologic parameters and operational decisions to engineering (CO₂ injection rates, plume dimensions, and storage capacities and effectiveness) and costs. Our results show, for the first time, benefits and impacts such as allowing CO₂ plumes to overlap, how different well spacing patterns affect CO₂ sequestration, the effects on costs of including brine treatment and disposal, and the effect of restricting injection rates to 1 MtCO₂/yr based on well limitations. We reveal multiple novel and unintuitive findings including: (i) deeper reservoirs have reduced carbon sequestration costs until injection rates reach 1 MtCO₂/yr, at which point deeper reservoirs become more expensive, (ii) thicker formations allow for increased injection rates and storage capacity, but thickness barely impacts plume areas, (iii) higher geothermal gradients result in reduced sequestration costs, unless brine treatment/disposal costs are included, at which point reservoirs having lower geothermal gradients are more economical because they produce less water for each unit of injected CO₂, and (iv) allowing plumes to overlap has a significantly positive impact of increasing storage capacities but has only a small influence on reducing sequestration costs. Overall, our results illustrate new scientific conclusions to help identify suitable sites to inject and store CO₂, to help understand the complex interaction between geology and resulting costs, and to help support the pursuit of meeting global sequestration targets.

KEYWORDS

CO₂ capture and storage (CCS); carbon sequestration; *SCO₂T*; reduced-order models; sensitivity analysis.

39 1.0 INTRODUCTION

40 Geologic carbon dioxide (CO₂) storage is a component of both CO₂-enhanced oil recovery (CO₂-EOR),
41 where CO₂ is injected into oil fields to increase energy production¹⁻³, and the CO₂ capture and storage (CCS)
42 process, where CO₂ from large point sources (e.g., coal-fired power plants, iron and steel manufacturing
43 facilities) is captured and compressed, transported in dedicated pipelines, and then sequestered in geologic
44 formations (e.g., deep saline aquifers, depleted shale gas formations)⁴⁻¹⁰ (Figure 1). It is also an essential
45 component of negative CO₂ emission technologies such as bioenergy coupled with CCS¹¹⁻¹³ (including
46 biorefineries and biomass power plants) or direct air capture¹⁴⁻¹⁶, which existing climate action plans
47 increasingly rely on to limit warming. Overall, geologic CO₂ storage is a critical technology for environmental,
48 energy, and economic security¹⁷ because it is part of every major climate action plan that limits warming to
49 below 2°C¹⁸⁻²⁰ and the beneficiary of substantial CO₂ tax credits in the United States (US) that are likely to
50 jumpstart a CO₂ storage industry at a large scale²¹⁻²⁴.

51 To meet environmental, energy, and economic goals, hundreds of millions or billions of tonnes of CO₂
52 annually (100s MtCO₂/yr to 1+ GtCO₂/yr) must be sequestered globally. For example, the
53 Intergovernmental Panel on Climate Change indicates the need to reduce emissions by more than 40
54 GtCO₂/yr by mid-century to achieve net-zero emissions and limit warming to 1.5°C, which some estimate
55 could require up to 300 GtCO₂ sequestered by 2050²⁵. While studies to date suggest that there is more than
56 enough subsurface storage resource to sequester hundreds of gigatonnes of CO₂—deep saline aquifers are
57 fairly ubiquitous worldwide²⁶, underlying approximately half of North America, for example²⁷⁻²⁹, and resource
58 estimates range between 2,400 and 3,700 GtCO₂³⁰ for the continental United States alone—the storage
59 potential of these geologic formations varies. Meeting the gigatonne challenge requires a significant advance
60 from estimating the prospective geologic storage resource of an area (e.g., US Department of Energy's
61 NATCARB database²⁸) to the geologic storage capacity based on explicit consideration of the dynamic
62 processes associated with CO₂ injection and migration. In addition, the geologic properties (e.g., permeability,
63 depth, thickness) have complex and nonlinear interactions that differentially affect the cost of CO₂ storage by
64 factors like the maximum CO₂ injection rate allowed by the formation and the dimensions of the CO₂ plume
65 in the subsurface. As a result, identifying formations and geographic locations to target for geologic CO₂
66 storage is not trivial and is a pressing challenge. This challenge is only intensified when design and operational
67 decisions, such as allowing CO₂ plumes in the subsurface to overlap (an option for increasing the density of
68 wells at a given CO₂ storage site that may decrease cost) or extracting and treating brine, are also considered.
69 This work aims to contribute to meeting the gigatonne challenge by providing a new and unique capability for
70 evolving from storage resource assessment to the required storage capacity assessment, including cost
71 estimates for CO₂ storage in deep saline formations.

72 In this study, we perform seven sensitivity experiments that are designed to identify, explore, and quantify
73 CO₂ sequestration properties of geologic formations and to understand why, for example, combinations of
74 geology and operational decisions lead to preferable outcomes such as lower costs, higher injection rates, or
75 higher storage capacities. We evaluate the sensitivity of CO₂ storage performance (rate of CO₂ injection,
76 plume radius, plume volume, storage area density, storage capacity, well density, and cost) to reservoir
77 formation properties (depth, thickness, permeability, porosity, and temperature) and reservoir operational
78 decisions (well spacing, brine extraction and treatment, and realistic injection capacities based on well-casing
79 diameters). The specific operational decisions we consider are (a) constraining CO₂ injection to 1 MtCO₂/yr

80 per well, which is a realistic maximum injection rate based on well-casing diameters, (b) the option to allow
81 CO₂ plumes in the subsurface to overlap, and (c) the effect of producing and treating/disposing of subsurface
82 brine, which was previously investigated as a means to minimize induced seismicity and leakage risk and
83 increase CO₂ storage potential³¹⁻³³. This extracted brine must be treated and/or disposed of, with a cost as
84 much as several dollars or more per cubic meter of brine³⁴. And, with a representative reservoir CO₂ density
85 of 666 kg/m³ producing around 1.5 cubic meters of brine for each tonne of CO₂ injected, brine treatment
86 and disposal can add several dollars per tonne to CO₂ sequestration. These operational and design decisions
87 are likely to have substantial ramifications for selecting and prioritizing site selection and, unlike subsurface
88 properties, they are variables that CO₂ storage operators and site planners can control.

89 The paper is arranged as follows: first, in the background section, we present a literature review of previous
90 sequestration analyses followed by a description of the framework we use, called *SCO₂T*³⁵, which is designed
91 to perform the sensitivity analyses required for this study. Second, we outline the methodological approach
92 used in our seven sensitivity experiments. Third, we present the results and discussion of the seven analyses.
93 Fourth and finally, we wrap up with conclusions and implications of the research including areas of further
94 exploration and development.

95 **2.0 BACKGROUND**

96 *2.1 Literature review*

97 Despite the importance of understanding the performance of geologic CO₂ storage to meet global climate
98 goals, the knowledge domain of the CO₂ sequestration science field of study is relatively nascent, in part
99 because appropriately investigating geologic CO₂ storage typically requires using full-physics numerical
100 simulations that are prohibitively resource-intensive. Several sensitivity analysis studies of CO₂ storage
101 performance to uncertain parameters, including storage formation properties and/or operational decisions,
102 have been conducted in recent years using existing tools. Here, we provide an overview of the most relevant
103 studies to the *SCO₂T*-based sensitivity analysis introduced in this paper.

104 McCoy and Rubin³⁶ performed a probabilistic analysis that quantified the sensitivity of CCS cost models to
105 variability in design parameters, including pipeline transport model parameters (e.g., pipeline length and
106 inlet/outlet pressure) and geologic parameters (e.g., reservoir permeability, depth, and thickness). Their results
107 showed that the cost of CO₂ injection highly depended on the permeability of the storage reservoir. They also
108 showed that the cost of CO₂ storage increases as reservoir permeability decreases, and the cost of both
109 transport and storage decrease with increasing power plant capacity factor. Stauffer et al.³⁷ developed a system
110 model called *CO₂-PENS* (Predicting Engineered Natural Systems) for evaluating the viability of CO₂ storage
111 at a range of sites. Monte Carlo simulations can be performed via *CO₂-PENS* to explore complex interactions
112 between uncertain parameters and distinguish the likely performance of potential storage sites. As an
113 example, Stauffer et al.³⁸ showed how the calculations of the number of wells required and the estimates of
114 plume size can affect long-term storage costs, including decreased costs with increasing reservoir depth.
115 Sifuentes et al.³⁹ analyzed the impact of various physical properties on the effectiveness of CO₂ sequestration
116 in aquifers. Experimental design-based sensitivity analysis was conducted to identify the most important
117 parameters for the trapping of CO₂. It was observed that horizontal permeability is the most impactful
118 parameter on the total amount of CO₂ dissolved into the brine, and residual gas saturation was found to be
119 the greatest contributor to the total amount of residual CO₂. Also, permeability heterogeneity is a major

120 contributor to both CO₂ trapping mechanisms. Later, Wainwright et al.⁴⁰ conducted sensitivity analyses based
121 on a basin-scale reservoir model developed for a hypothetical storage project located in the southern San
122 Joaquin Basin in the US. The impact of uncertainty in parameters (e.g., reservoir permeability, porosity, and
123 pore compressibility) on risk-related performance measures, including the CO₂ plume area and pressure
124 plume size, was evaluated. Three different sensitivity analysis methods—a local sensitivity method, the global
125 Morris method, and the Sobol/Saltelli method—were compared. Results showed that the three analysis
126 methods provided identical interpretations and importance rankings, and that reservoir permeability was
127 identified as the most important parameter for all the performance measures. Metcalfe et al.⁴¹ developed a
128 generic system model using Quintessa's QPAC software, and the model was then applied to the evaluation of
129 CO₂ storage performance at Krechba, near In Salah in central Algeria. Although sensitivity analysis of overall
130 system performance to key operational decisions in the system can be made, the sensitivity analysis of the
131 storage performance including well density and costs to reservoir geologic properties cannot be conducted
132 using their model.

133 The sensitivity analysis of CO₂ storage performance on uncertain parameters has also been carried out for
134 storage at CO₂-enhanced oil recovery (CO₂-EOR) sites. Dai et al.^{42,43} quantified the sensitivity of a set of risk
135 metrics, including CO₂ injection rate, cumulative CO₂ storage, and the total amount of oil production to
136 uncertain parameters such as reservoir thickness, depth, permeability, and porosity. The results showed that
137 the CO₂ injection rate and the amount of CO₂ storage are most sensitive to reservoir permeability, thickness,
138 and injection pressure. Cumulative oil production is mainly controlled by well spacing, reservoir pay zone,
139 permeability, and initial oil saturation. Chen and Pawar² identified key geologic and operational characteristics
140 that affect CO₂ storage capacity and oil recovery potential by performing Monte Carlo simulations and
141 sensitivity analyses. They found that cumulative CO₂ injection is mainly controlled by reservoir permeability
142 and that the total amount of CO₂ retained in the reservoir increases with increased producer bottom-hole
143 pressure, reservoir thickness, and CO₂ injection rate, while it decreases with increased reservoir permeability.

144 A few system models or tools were developed by researchers to evaluate CO₂ storage performance, but the
145 system models or tools developed in their work cannot perform sensitivity analysis to show how uncertain
146 geologic parameters and/or operational decisions affect the CO₂ storage performance (e.g., injection rate,
147 plume radius, well density, and cost). For example, Zhang et al.⁴⁴ developed a system-level model based on
148 *GoldSim* for evaluating CO₂ storage performance in a gas reservoir. However, as mentioned by the authors in
149 their conclusion, the sensitivity analysis of CO₂ storage performance on uncertain geologic/operational
150 parameters could not be performed with their model.

151 Although previous work examined the sensitivity of carbon sequestration to different geologic parameters
152 and ranges, a comprehensive analysis of how major geologic parameters and operational decisions affect CO₂
153 storage performance, including both engineering (CO₂ injection rates, plume dimensions, and storage
154 capacities and effectiveness) and costs, has not yet been conducted. This key science gap exists because none
155 of the previously discussed tools, methods, or models can quickly analyze thousands of dynamic reservoir
156 simulations for a single site, or millions over many sites, to identify ideal reservoir properties and the
157 sensitivity and interactions between geologic parameters and operational decisions. Understanding these
158 impacts and complex feedbacks requires a tool that can run while still being able to capture key sequestration
159 processes (i.e., real dynamics of injecting and storing CO₂). That is, the tool should combine the power of
160 full-physics numerical simulations, which are prohibitively resource-intensive, with the speed of systems-level
161 tools (which lack the required physics).

162 *SCO₂T* is a systems tool that was originally developed to support sink characterization for the *SimCCS*
163 framework^{45, 46}, but can be applied to this challenge. Previous *SCO₂T* publications have discussed the *SCO₂T*
164 software and used it to demonstrate well-known sequestration relationships, but did not address these larger
165 knowledge gaps in the sequestration science field³⁵.

166 **2.2 *SCO₂T***

167 *SCO₂T* uses reduced-order models (ROMs) to replicate key outputs from the Finite Element Heat and Mass
168 (FEHM⁴⁷⁻⁴⁹) Transfer Code that simulates complex multi-fluid/multi-phase fluid flow (in this case, CO₂ and
169 water). The use of ROMs allows it to maintain nonlinear feedbacks and interactions while being able to
170 simulate thousands of scenarios per second. The tool is able to accurately simulate dynamic CO₂ injection
171 rates and CO₂ plume dimensions (from the full-physics simulator FEHM simulator), using the ROM
172 approach (called Frankenstein's ROMster⁵⁰) and couple outputs with detailed sequestration economics
173 including injection, storage, and brine treatment costs.

174 It should be emphasized that, even though *SCO₂T* uses perhaps the most detailed economic inputs from the
175 US Environmental Protection Agency (EPA)⁵¹ in a similar approach to the FE/NETL CO₂ Saline Storage
176 Cost Model⁵², these costs are still very much uncertain. In reality, they could vary from project to project, one
177 key reason the model includes an uncertainty option. Many costs can also be varied by the user, allowing for
178 uncertainty in storage costs due to project variance to also be addressed. However, even though the impact
179 on costs in our seven experiments is perhaps within the range of uncertainty, it is still illustrative to examine
180 the relative effect on costs. *SCO₂T* assumes the maximum possible injection rate up to 80% of lithostatic
181 pressure. Capturing dynamic plume dimensions is particularly important because it is critical to understanding
182 how efficiently CO₂ can access the pore space and how an operator might spatially arrange injection wells.
183 The model also allows a fractional-well option where a user can allow a non-integer number of wells. This
184 option is useful for sensitivity analysis, where the user doesn't want the model to increase costs in a saw-tooth
185 pattern as new integer wells are added, or to enable direct comparison between different sites particularly with
186 small 2D footprints. See Figure 1 for how plume dimensions can change with, for example, increasing depth.

187 Although *SCO₂T* has made many advances to enable this current study, particularly capturing dynamic CO₂
188 injection and storage and linking this to economics, there are many important sequestration aspects that it
189 does not try to simulate (and are not part of this study), many which are available in the models mentioned in
190 the literature review. For example, the model doesn't simulate processes such as wellbore leakage, induced
191 seismicity, or hydrocarbon production. Further, *SCO₂T*, like similar ROM-based approaches, makes key
192 reservoir and modeling assumptions, including lithostatic and hydrostatic pressures, residual water fraction,
193 endpoint relative permeability, more complex geologies, and other parameters including heterogeneity,
194 different fluid properties, and different depositional environments. The current version also does not include
195 monitoring, post-injection, and site closure costs; these will be developed in future versions, along with
196 addressing other model updates, including the ability to simulate injection between 5 and 50 years.

197 **3.0 APPROACH**

198 We use the *SCO₂T* framework to explore how reservoir properties (depth, thickness, permeability, porosity,
199 and temperature) and operational decisions (maximum injection rate, well spacing, and brine treatment)
200 impact the ability to inject and store CO₂ and the final sequestration cost. We analyze how these variables
201 affect CO₂ characteristics that control the injection rate and CO₂ plume dimensions including (i) CO₂ density

202 (kg/m³) and (ii) CO₂ viscosity (Pa*s). CO₂ density and viscosity principally control how easily CO₂ can be
203 injected and how easily CO₂ can spread out in a reservoir in three dimensions. Consequently, these properties
204 impact key storage parameters including (iii) maximum CO₂ injection rates (MtCO₂/yr), (iv) CO₂ plume radii
205 (km), (v) CO₂ plume volumes (million cubic meters; Mm³), and (vi) storage area density (MtCO₂/km²).
206 Storage area density is a measure of storage effectiveness for the 2D footprint of available land and is based
207 on a single non-overlapping CO₂ plume.

208 We perform a one-at-a-time (OAT) or transect sensitivity analysis across the five key geologic variables⁵⁰ to
209 understand the complex interaction of the variables and their impact on sequestration engineering and
210 economics. We also apply a sensitivity analysis to non-geologic and operational variables that have a
211 significant impact on economics including assigning increasing costs to produced brine (brine can be
212 produced to minimize induced seismicity risks and to increase CO₂ storage potential) and allowing CO₂
213 plumes to overlap by placing CO₂ injection wells increasingly close together.

214 The paper consists of seven OAT sensitivity analyses that explore the impact on CO₂ injection and storage by
215 varying (1) “artificial” depth through manipulating pressure and temperature independently and jointly, (2)
216 actual depth, (3) reservoir thickness, (4) permeability, (5) porosity, (6) geothermal gradient and brine
217 treatment, and (7) injection-well spacing (overlapping CO₂ plumes). These seven experiments were chosen to
218 cover perhaps the most important range of variable geologic parameters (five) and operational decisions
219 (three); exploration of the geothermal gradients and brine treatment are covered in a single experiment, hence
220 seven experiments. For each experiment, the geologic parameters/ranges and assumptions were chosen to
221 represent realistic parameters from existing studies and field sites while ensuring that critical operational
222 impacts were represented. For example, the impact of limiting injection rates to 1 MtCO₂/yr can only be
223 shown for an experiment with geologic variables, such as permeability, that allow this injection rate. Values
224 for the five key geologic variables are listed in the figures for each experiment.

225 4.0 RESULTS AND DISCUSSION

226 The color schemes used in charts for each experiment are consistent throughout. For example, from Figure 2,
227 CO₂ density is blue and CO₂ viscosity red throughout all figures in the paper. Charts with dotted lines refer to
228 where the maximum injection rate was limited to 1 MtCO₂/yr. Site- and economic-related outputs, including
229 reservoir storage capacity, number of wells, and final sequestration costs are calculated only using the rate-
230 limited simulations (i.e., when the rate is limited to 1 MtCO₂/yr by the injection well engineering limits); it
231 makes no sense to calculate reservoir economics for a reservoir that could theoretically inject 10 MtCO₂/yr in
232 a single well because typical injection wells will be limited to around 1 MtCO₂/yr. All data for every
233 experiment and figure is available in an Excel Workbook in the supporting information (SI), including the
234 input data to run the *SCO₂T* model and all relevant outputs. The results for the first two experiments—
235 exploring temperature and depth independently and together—are described and discussed in greater detail
236 than the remaining five experiments to avoid repetition in later experiments.

237 4.1 Experiment #1: Pressure, Temperature, and Depth

238 The first experiment evaluates the impact of formation depth on CO₂ properties and CO₂ sequestration
239 potential. As formation depth can generally be considered positively correlated with pressure and
240 temperature, we use pressure and temperature as a proxy for the formation depth in parts of the analysis.
241 Pressure increases with depth because of the overlying rock and water mass, while temperature increases

242 owing to Earth's geothermal gradient. Figure 2 illustrates the impact of increasing pressure while holding
243 temperature constant (three panels in the top row), increasing temperature while keeping the pressure
244 constant (middle row), and simultaneously increasing pressure and temperature (bottom row). The input data
245 for Figure 2 is provided in the Excel Workbook in the SI.

246 The formation pressures considered in the top row of Figure 2 range from 9.8 MPa to 29.4 MPa (primary x-
247 axis), which coincides with the formation depth ranging from 1000 to 3000 m (secondary x-axis).
248 Temperature is held at 56°C to isolate the impact of pressure. At 1000 m, 56°C translates into a geothermal
249 gradient of 45°C/km (maximum geothermal gradient plus an 11°C ground surface temperature—this ground
250 surface temperature is used on all simulations in this study) and at 3000 m 56°C translates to a geothermal
251 gradient of 15°C/km (minimum geothermal gradient for *SCO₂T*). The formation temperatures considered in
252 the middle row of Figure 2 correspond to geothermal gradients from 15 to 45°C/km at a depth of 2000 m.
253 This translates into a temperature range of 41 to 101°C. These simulations isolate the impact of temperature
254 on CO₂ properties, injection rates, and plume properties while keeping the pressure constant. The formation
255 depths considered in the bottom row of Figure 2 range from 1000 to 3000 m while letting pressure vary
256 according to the hydrostatic gradient and temperature vary according to a 25°C/km geothermal gradient.
257 (*SCO₂T* also allows users to independently input pressure and temperature based on known values.) The
258 default reservoir hydrostatic pressure is calculated as a function of depth by assuming an average water
259 density of 1000 kg/m³ with the acceleration of gravity as 9.807 m/s². This translates into 9.807 MPa of
260 pressure increase per kilometer of depth.

261 The pressure and temperature changes associated with increasing depth have opposite impacts on CO₂
262 properties and sequestration characteristics. Increasing pressure-only (Figure 2abc) increases both the CO₂
263 density and viscosity (Figure 2a; blue and red lines, respectively). Increased density means that more CO₂ can
264 be stored in the available pore space, increasing the storage area density (Figure 2b; orange line). Increased
265 viscosity typically would lead to a reduced CO₂ injection rate (viscous fluids flow less easily), however, the
266 rate of increase in CO₂ density (which increases injection rates) overpowers the impact of viscosity.
267 Consequently, CO₂ can be injected at a higher rate and thus a greater mass of CO₂ can be injected in a well
268 (Figure 2b; green line). Although increasing CO₂ density means that the plume radius would decrease for a
269 constant injection rate, the increased CO₂ mass (owing to the increased injection rate) overrides this effect
270 and so the plume radius increases with rising pressure (Figure 2c; red line). The same is largely true for plume
271 volume—here, plume volume is the volume of the CO₂ only (equivalent to the injected volume) and not the
272 volume associated with the region of the reservoir where CO₂ can be found in the pore space—where the
273 increasing rate of CO₂ injected overrides the increasing CO₂ density (Figure 2c; purple line). The falling
274 plume volume between 9.8 MPa and 13 MPa is due to the competing effects between CO₂ density and
275 viscosity which both have large rates of change in this range. In *SCO₂T*, the plume volume is only used to
276 estimate the volume of brine that is displaced by the injected CO₂, thus it doesn't consider the rock volume
277 that the plume occupies (i.e., 1:1 relationship between CO₂ plume volume and displaced brine in reservoir
278 conditions). Future versions will potentially generate an actual plume volume ROM.

279 Increasing temperature-only (Figure 2def) reduces both CO₂ density and viscosity (Figure 2a). Here, the
280 injection rate rises with increasing depth (Figure 2e); the reduced CO₂ density would be expected to reduce
281 the maximum injection rate, but this is overridden by the reduced viscosity of CO₂ and how it influences flow
282 in porous media. Increasing temperature has a more complex impact on the storage area density (Figure 2e).
283 Up to 66°C the storage area density is relatively constant at around 0.4 MtCO₂/km² (66°C corresponds to a

284 geothermal gradient of $27.5^{\circ}\text{C}/\text{km}$ and 11°C for the ground surface temperature) because the impacts of CO_2
285 density and the larger mass of CO_2 injected balance each other. Beyond 66°C , the storage area density falls
286 because the injection rate does not continue to rise quickly enough to counterbalance the reduced CO_2
287 density. The plume radius and plume volume both rise with increasing temperature (Figure 2f); increased
288 injection rates and reduced CO_2 density increase both the volume of the plume and how far the plume
289 spreads. The CO_2 injection rate never reaches the $1 \text{ MtCO}_2/\text{yr}$ maximum.

290 Figure 2ghi illustrates the impacts when pressure and temperature are both allowed to increase simultaneously
291 as depth goes from 1000 to 3000m. Pressure (Figure 2a) and temperature (Figure 2d) have opposing impacts
292 on CO_2 density and viscosity, which are nearly balanced out when pressure and temperature vary naturally
293 with depth (Figure 2g). However, the overriding impact of CO_2 density means that both the injection rate and
294 storage area density both rise with increasing depth (Figure 2h); in this case, the impact of pressure has a
295 much greater impact than temperature. In the case of the plume radius and volume, increasing depth forces
296 both to rise (Figure 2i). Interestingly, despite their counterbalancing effects, the combined impacts of pressure
297 and temperature lead to both higher (in deeper formations) and lower (in shallower formations) values for
298 CO_2 injection rates and storage area capacities (Figure 2h) and plume radii and volumes (Figure 2i) than that
299 for pressure (Figure 2bc) and temperature (Figure 2ef) independently.

300 This first experiment illustrates that there is a complex and nonlinear interaction between pressure and
301 temperature in terms of CO_2 properties and sequestration characteristics. Being able to capture these effects
302 is critical for understanding sequestration, particularly when coupled with economics. This further
303 demonstrates the need for reduced-order modeling approaches to be able to capture effects that are typically
304 captured only when using full-physics numerical models but with the speed of analytical models.

305 **4.2 Experiment #2: Depth**

306 The second experiment is a more traditional exploration of the impacts of increasing depth on sequestration
307 engineering and economics, where formation depth is increased from 1000 to 5000 m while holding the
308 remaining four geologic variables constant (Figure 3). The solid lines in Figure 3a and Figure 3b represent
309 simulations where the injection rate is unlimited, whereas the dotted lines represent simulations where the
310 injection rate is limited to the $1 \text{ MtCO}_2/\text{yr}$ maximum rate governed by well-casing diameters. For the
311 unlimited runs, the injection rate and plume radius both rise with increasing depth, though the rate of increase
312 in the plume radius slowly declines (Figure 3a). Consequently, the plume volume and storage area density for
313 the unlimited simulations also rise with increasing depth (Figure 3b). Although the plume area radius
314 increases, the storage density increases faster than the expanding plume radius (the CO_2 plume expands at the
315 bottom of the formation faster than the spread at the top of the formation; see the Thickness Experiment
316 below for more details). In this experiment, the rate of increase in CO_2 density caused by the pressure
317 increase overrides the decrease in CO_2 density by temperature. Consequently, more CO_2 can be injected with
318 depth and more effectively or more densely occupies the available pore volume given the parameters in this
319 experiment.

320 Limiting the injection rate to $1 \text{ MtCO}_2/\text{yr}$ imparts significant changes (Figure 3a and Figure 3b). The 1
321 MtCO_2/yr injection rate is reached at a depth of 2850 m (see data in the SI for Figure 3). At this threshold,
322 the simulated injection rate is capped at exactly $1 \text{ MtCO}_2/\text{yr}$ (Figure 3a; green dotted line). However, the rate-
323 limited plume area linearly declines after the 2850 m depth is reached (Figure 3a; red dotted line). That is, if
324 you inject a constant amount of CO_2 , the distance the CO_2 migrates laterally decreases as the formation depth

325 increases because that same amount of CO₂ occupies a smaller 3D space (i.e., is denser). Put another way, the
326 plume area in the unlimited case continues to expand because the increasing rate of CO₂ injected overrides
327 the impact of more-densely stored CO₂. In our case, the CO₂ plume volume—the volume of the injected
328 CO₂—remains relatively steady (but not flat) after 2850 m because the competing effects of pressure and
329 temperature keep the CO₂ density relatively flat (Figure 3b; dotted purple line). After the 1 MtCO₂/yr
330 threshold is reached, the storage area density continues to rise but not quickly as in the unlimited case (Figure
331 3b; orange dotted line). This is because the plume radius falls in the rate-limited case and thus, in terms of 2D
332 storage area density, the storage effectiveness still increases even though the injection rate is held at 1
333 MtCO₂/yr.

334 The bottom panes in Figure 3 (Figure 3c and Figure 3d) illustrate the logistic and economic sequestration
335 impacts. The bottom two panels in Figure 3, Figure 4, Figure 6, and Figure 7 are all based on rate-limited
336 simulation results. Storage capacity rises with increasing depth (Figure 3c; olive-colored line); all simulations
337 assume a 256 km² (16 x 16 km) surface footprint for a reservoir. The storage capacity of the reservoir is
338 closely related to the storage area density. The storage area density is based on a single CO₂ plume within a
339 fixed area, while the reservoir storage capacity is based on injecting CO₂ into multiple wells within a fixed area
340 considering the space between plumes and areas of plume overlap. The number of wells that can be placed in
341 the 16 x 16 km area decreases as thickness increases (because the CO₂ plume 2D footprints get larger and
342 larger) until the point when the injection rate reaches the 1 MtCO₂/yr threshold (Figure 3c; light blue line).
343 The number of wells is calculated by working out how many injection plumes could be placed in the available
344 area, accounting for the spatial arrangement of wells and the amount of allowable overlap. In all cases apart
345 from the Well Spatial Arrangement Experiment, all wells are assumed to be placed in a hexagonal pattern
346 with no plume overlap. Further, to ensure that the sensitivity analysis is not affected by the integer nature of
347 wells, we use the *SCO₂T* fractional well option. Once the 1 MtCO₂/yr threshold is reached, the plume radius
348 falls and the number of wells that can be placed in the available area rises. The number of wells that are
349 required to maximize the injection rate in the 16 x 16 km area and the injection rate of each well are the two
350 key parameters for calculating overall sequestration costs (Figure 3d; teal line). *SCO₂T* has customizable
351 options that exogenously impact costs: whether to include costs to pump CO₂ (no pumping is required if
352 CO₂ arrives at the wellhead at suitable pressure), cost to drill brine production wells, and cost to treat brine.
353 Here, in all cases apart from the Geothermal Gradient and Brine Treatment Experiments, these costs are set
354 to zero. Down to the 2850 m (or 1 MtCO₂/yr) threshold, sequestration costs decline because fewer wells are
355 required *and* more CO₂ can be injected in each well (site-wide fixed costs, such as 3D seismic surveys and site
356 preparation, are constant). However, beyond the threshold, the costs rise because, even though storage area
357 density still increases, the storage effectiveness of each well declines.

358 This last point is critical for understanding the suitability of sequestration characteristics; looking at injection
359 rates or storage area density or storage capacity alone is not a good indicator of sequestration suitability. Here,
360 imagine comparing two hypothetical reservoirs where all formation properties are identical except that one
361 has a depth of 3000 m and the other 5000 m. The latter has a higher injection rate, higher storage area
362 density, and a higher total reservoir capacity, yet its costs are more than 10% higher (\$4.35/tCO₂ versus
363 \$3.92/tCO₂). This is a key attribute of *SCO₂T* and how the framework captures the complexity and fidelity of
364 sequestration engineering (e.g., limited and unlimited injection rates, plume dimensions) and links it to
365 sequestration economics based on the actual logistics of injecting and storing CO₂.

366 **4.3 Experiment #3: Thickness**

367 The third experiment examines the impact of increasing formation thickness (Figure 4). As thickness
368 increases from 5 to 100 m, the unlimited CO₂ injection rate, in this case, climbs from 0.14 MtCO₂/yr to
369 almost 2 MtCO₂/yr (Figure 4a). The unlimited CO₂ injection rate, again, is simply the case where the injection
370 rate is not limited to the maximum 1 MtCO₂/yr rate enforced by the well engineering limits. This is an almost
371 15-fold increase in the injection rate as the thickness increases by 20 times. At the same time, the CO₂ plume
372 radius is almost flat, with only a slight decrease. That is, even though the CO₂ injection rate increases by 15
373 times, the plume radius is almost unchanged. Although the theoretical shape of CO₂ plumes was investigated
374 previously (e.g., Nordbotten et al.⁵⁵), this effect was unexpected and has perhaps not been clearly documented
375 in the literature. Here, although the mass (and volume) of CO₂ massively increases, the available pore space
376 increases at a faster rate and the plume does not have to spread out as far. Figure 5 illustrates this unexpected
377 result through a series of full-physics FEHM-simulated CO₂ plumes in formations with varying formation
378 thickness that correspond to the simulations in Figure 4. When the injection rate is limited to 1 MtCO₂/yr
379 (occurring at a thickness of 46 m) the CO₂ plume radius reduces with increasing thickness (Figure 4a; dotted
380 red line) because the injected CO₂ mass is constant, whereas the available pore volume continues to increase.
381 *SCO₂T* assumes that the screen size—the vertical extent of the wellbore where CO₂ can pass into the
382 formation—stretches across the entire formation thickness. The CO₂ plume volume expands correspondingly
383 with the limited and unlimited injected CO₂ mass (no changes in CO₂ density and temperature), while the
384 storage area density rises with increased thickness (Figure 4b). However, the storage area density for the
385 limited injection rate is not quite as high as for the unlimited injection rate (i.e., where the injection rate can
386 be greater than 1 MtCO₂/yr), indicating that the effect of increased thickness outweighs the effect of the
387 decreased CO₂ plume radius.

388 The reservoir storage capacity increases correspondingly with the rising storage area density (Figure 4c). The
389 number of wells corresponds with the plume radius based on the limited injection rate; once the 1 MtCO₂/yr
390 kicks in at 46 m, the number of wells increases because the footprint of the plumes is smaller and thus more
391 wells can be packed together. The rapidly increasing injection rate and storage area density coupled with the
392 constant number of wells (up to a thickness of 45 m) means that the sequestration cost drops by more than
393 50%—\$11.19/tCO₂ to \$4.93/tCO₂ from 5 to 45 m (Figure 4d). Beyond a thickness of 45 m, the
394 sequestration costs are flat because the increased cost of using more wells is balanced out by the increased
395 storage area density.

396 **4.4 Experiment #4: Permeability**

397 The fourth experiment focuses on the impact of increasing permeability (Figure 6). Permeability is well-
398 known to affect injection rates which, in turn, has a significant impact on sequestration costs. Figure 6a shows
399 injection rates rapidly increasing as permeability rises from 1 to 150 mD, which in turn increases the CO₂
400 plume radius (Figure 6a). When the injection rate is capped at 1 MtCO₂/yr (at 82 mD in the experiment), the
401 CO₂ plume radius continues to expand (though less quickly than the unlimited injection case) because the
402 increased permeability allows the CO₂ to migrate laterally more easily. The rate of increase of the CO₂ plume
403 radius gradually declines between 1 and 60 mD before reaching a near-steady rate of increase (Figure 6a). The
404 CO₂ plume volume (i.e., the volume of brine that is displaced by the injected CO₂) is directly related to the
405 unlimited/limited CO₂ injection rate (compare the green line in Figure 6a with the purple line in Figure 6b).
406 The storage area density increases between 1 and 60 mD (Figure 6b), corresponding to the part of the CO₂

407 plume radius that has a decreasing rate of increase (Figure 6a). Beyond 60mD, the storage area density is flat
408 for the unlimited injection rate because the effect of injected CO₂ mass and plume radius on storage area
409 density cancel each other out. In the limited injection case (i.e., beyond 82 mD) the storage area density
410 declines because the injection rate is constant but the CO₂ plume radius continues to increase.

411 Because the plume radius continues to expand for the limited injection rate (and unlimited also), the number
412 of wells decreases as permeability rises from 1 to 150 mD (Figure 6c). The reservoir storage capacity follows
413 the same general trend as the storage area density: the capacity rises between 1 and 60 mD, is flat until the 1
414 MtCO₂/yr threshold is reached (at 82 mD), and then the reservoir storage capacity falls (even though each
415 well has the same injection rate, their plume radii are larger). These effects combine for a nonlinear impact on
416 sequestration costs (Figure 6d). Sequestration costs drop exponentially from around \$40/tCO₂ at 1 mD
417 (injection rate of 0.015 MtCO₂/yr) to around \$6.50/tCO₂ once permeability rises to 20 mD (injection rate of
418 0.25 MtCO₂/yr). This exponential decline in costs as injection rates rise to around 0.25 MtCO₂/yr has been
419 shown previously⁵⁴. Subsequently, sequestration costs keep falling to \$4.64/tCO₂ at a permeability of 82 mD
420 before slowly rising to \$4.92 at 150 mD. The steady increase in costs after 82 mD is caused by the decline in
421 storage area density (i.e., storage effectiveness).

422 **4.5 Experiment #5: Porosity**

423 The fifth experiment focuses on porosity (Figure 7) and is the simplest in the sense that porosity does not
424 affect CO₂ density, CO₂ viscosity, or the movement of CO₂. Consequently, the CO₂ injection rate is flat as
425 porosity moves from 0.05 to 0.4 (Figure 7a). Although porosity does not affect the movement of CO₂ in our
426 experiment, the plume radius does decline because a greater volume of pore space can be filled in all
427 directions. A linear change in porosity does not have a linear effect (decline) on the plume radius because the
428 2D radius is a function of the 3D plume shape or volume. Because the injection rate does not change, the
429 CO₂ plume volume does not change while the storage area density increases along with the linear increase in
430 pore volume (Figure 7b). Consequently, both the reservoir storage capacity and the number of wells increases
431 nearly linearly with porosity (Figure 7c). At first, it seems intuitive that reservoirs having higher porosities
432 would require more wells, but it becomes evident when considering that the plume radii do decrease with
433 increasing porosity. Although the number of wells rises, the effect of the linear increase in storage area
434 density overcomes the effect of the increase in the number of wells on total sequestration cost, which declines
435 with increased porosity (Figure 7d).

436 **4.6 Experiment #6: Geothermal Gradient and Brine Treatment**

437 The sixth experiment explores the effect of geothermal gradient with and without the costs of treating
438 produced brine (Figure 8). Note that, although the effect of an increasing geothermal gradient on CO₂
439 injection, plume dimensions, and storage area density has already been described in the first experiment
440 (Figure 2d, Figure 2e, and Figure 2f), the bounding geologic parameters are slightly different in experiment
441 #6. As noted previously, extracting brine is likely to be critical for large-scale sequestration as a means to
442 minimize induced seismicity, minimize CO₂ leakage risk, and increase CO₂ storage potential, and that treating
443 and disposing of this brine will lead to additional sequestration costs. In *SCO₂T*, the user inputs their desired
444 \$/m³ cost to treat and dispose of brine.

445 In this experiment, the geothermal gradient increases from 15°C/km to 45°C/km—formation temperature
446 rises from 41°C to 101°C using a constant depth of 2000 m—the CO₂ injection rate rises in linear steps or

447 pieces and the plume radius rises almost linearly (Figure 8a). The rising injection rate and falling CO₂ density
448 lead to a rapidly increasing plume volume (Figure 8b). Because the injection rate increases fastest up to 60°C,
449 the storage area density happens to rise with increasing temperature (Figure 8b). Once the rate of increase in
450 the injection rate reduces slightly, between 60°C and 78°C, the storage area density declines slowly (i.e., a
451 tipping point). Then once the rate of increase in the injection rates falls more steeply (after 78°C), the storage
452 area density notably declines. However, it should be noted that the storage area density varies only between
453 around 0.27 MtCO₂/km² and 0.29 MtCO₂/km² as the geothermal gradient is changed from 15°C/km to
454 45°C/km because the injection rate and plume radius increase very similarly.

455 The number of wells falls almost linearly (Figure 8c) as a result of the increasing plume radius, dropping from
456 4.5 wells to 2.8 wells. The reservoir storage capacity (Figure 8c) follows the same pattern as the storage area
457 density plot, with the total available capacity varying no more than a few percent as the geothermal gradient
458 increases from 15°C/km to 45°C/km. That is, even though the number of wells that can be placed within the
459 2D area falls, the increased injection rate for each well compensates and so storage capacity is not significantly
460 affected. The final result for the simulations with no brine treatment costs (bottom curve in Figure 8d) is that
461 sequestration costs fall with the increasing geothermal gradient; total sequestration costs fall from \$3.55/tCO₂
462 (15°C/km) to \$3.07/tCO₂ (45°C/km). The CO₂ density drops from 829 kg/m³ (geothermal gradient of
463 15°C/km) to 466 kg/m³ (45°C/km), illustrating that more brine is produced as the geothermal gradient
464 increases (CO₂ density is not shown in the chart). *SCO₂T* assumes that the CO₂ displaces an equal volume of
465 brine. That is, even though the plume area expands (ultimately reducing storage area density), each well can
466 inject more CO₂ with the increasing geothermal gradient, and costs ultimately fall. However, assigning a cost
467 to treating brine changes this relationship, where, ultimately, higher geothermal gradients result in increased
468 costs (top two lines in Figure 8). That is, the cost to treat the increased brine production outweighs the
469 savings from injecting more CO₂. For example, a brine treatment cost of \$1/m³ of brine (middle lines in
470 Figure 8) sees costs fall slightly between 41°C to 61°C (\$4.74/tCO₂ to \$4.64/tCO₂) before rising to
471 \$5.14/tCO₂. With \$2/m³ brine treatment cost, sequestration costs rise almost continuously from \$5.96/tCO₂
472 to \$7.21/tCO₂. Consequently, sequestration costs having a \$2/m³ brine treatment cost are between 1.68 times
473 (15°C/km) to 2.35 times (45°C/km) higher than if brine is not produced or treated. A more than doubling of
474 sequestration costs could have a significant impact on sequestration economic analysis.

475 **4.7 Experiment #7: Well Spatial Arrangement**

476 Experiment #7 explores the impacts of arranging the injection wells in different patterns, focusing on the
477 effect of allowing plumes to overlap. At the outer edges of the plume, the CO₂ density is very low, and so
478 plume interaction is relatively minor with relatively low or no negative impacts. Figure 9 illustrates the
479 concept of overlapping plumes and how the same number of plumes (nine in this visualization) can be fit into
480 a smaller area when plume overlap is allowed. Consequently, the storage area density is increased as the pore
481 space is used more effectively. This experiment uses the overlapping plume capability to explore the effect of
482 allowing plumes to overlap between 0% and 50% (Figure 10). Note, the tool is a sequestration screening or
483 feasibility tool based on individual injection wells and does not attempt to capture the pressure interaction
484 between multiple plumes; instead, it assumes that pressures are managed through brine extraction. For an
485 experiment with a 50% overlap, this means that 50% of the area (not radius) of a plume is overlapped by
486 surrounding plumes, where the outer parts of the plume's horizontal extent have a lower storage area density
487 (i.e., thinner layer of CO₂) than closer to the center of the plume. Here, varying the plume overlap between 0
488 and 50% increases the storage capacity from 58 MtCO₂ to 97 MtCO₂ (Figure 10a; olive line), an increase of

489 67%, while the storage area density rises from 0.29 MtCO₂/km² to 0.39 MtCO₂/km² (orange line in Figure
490 10a) an increase of 35%. Allowing plumes to overlap means that more wells can be placed in the same 2D
491 area. Consequently, the number of wells rises from 3.4 to 5.7 (Figure 10b; light blue line), a rise of 28%.
492 Although having more wells translates into greater fixed capital and fixed O&M costs, more CO₂ can be
493 stored in the same 2D footprint and so overall sequestration costs fall from \$6.13/tCO₂ to \$5.75/tCO₂
494 (Figure 10b; dark blue line). However, this represents only a 6% drop in costs, which is essentially a function
495 of the fixed site-wide costs (e.g., purchasing land/pore space, permitting, seismic imaging) being spread over a
496 greater amount of stored CO₂. That is, costs directly related to drilling and operating each well do not change
497 by letting plumes overlap. A way to visualize this is to examine the 2D cross-profiles of potential plumes in
498 Figure 5; even with a high allowable overlap, the parts of the plume with a high storage density never overlap.

499 5.0 CONCLUSIONS AND IMPLICATIONS

500 Meeting global climate challenges will require identifying CO₂ sequestration sites on the scale of hundreds of
501 billions of tonnes of CO₂, which is orders of magnitude larger than current CO₂ sequestration levels. For
502 example, worldwide CO₂-enhanced oil recovery infrastructure currently sequesters less than 100 MtCO₂/yr,
503 and anthropogenic storage of CO₂ is around 40 MtCO₂/yr⁵⁵. In an effort to investigate the impact that
504 subsurface parameters and operational decisions have on the suitability of geologic CO₂ storage at this scale,
505 this study performed a set of sensitivity analyses using *SCo₂T*. The tool was used because it was designed to
506 be able to quickly capture complex and nonlinear interactions between geologic parameters—formation
507 depth, thickness, permeability, porosity, and temperature—and properties such as density and viscosity, and
508 how this affects CO₂ injection rates, plume dimensions, storage capacity, and overall economics. As such, the
509 primary finding of this study is the demonstration that it is possible to perform this required sensitivity
510 analysis to understand sequestration, which was up to now was a daunting and perhaps prohibitive task to
511 undertake using full-physics numerical models.

512 Our results and discussion are intended to help guide the emerging challenge of identifying suitable storage
513 sites on a massive scale. Our new findings are grouped into four sets of unexpected or previously unknown
514 conclusions:

- 515 1. *Limiting injection rates to a realistic 1 MtCO₂/yr per injection well has substantial impacts on sequestration engineering*
516 *and economics, even to the degree that presumed ideal geologic conditions could become less favorable.* For example, results
517 show that while identifying reservoirs with larger thicknesses (Experiment #3; Figure 4) and higher
518 porosity (Experiment #5; Figure 7) is typically going to result in improved sequestration, deeper
519 reservoirs (Experiment #2) and higher permeability reservoirs (Experiment #4; Figure 6) can have
520 poorer sequestration performance once the 1 MtCO₂/yr rate is reached. In the case of permeability, this
521 effect is relatively minimal, but is significant for depth: this suggests that using deeper reservoirs reduces
522 carbon sequestration costs until injection rates reach 1 MtCO₂/yr (engineering-limited rate of injection),
523 at which point deeper reservoirs cost more. This rate-limiting theme has the potential to have a
524 significant impact on how we reach aggressive sequestration targets.
525
- 526 2. *Thicker formations allow for larger injection rates and storage capacity but not plume area.* The impact of thickness on
527 sequestration and economics was partially unexpected and potentially significant for large-scale CO₂
528 injection and storage. Our results show that the radius of the CO₂ plume remains relatively constant—
529 which was unexpected and not previously documented—and may even decrease, with increasing
530 reservoir thickness. This occurs even though increasing thickness also increases the CO₂ injection rate as

531 well as the total CO₂ storage capacity (Experiment #3; Figure 4). This unexpected relationship is itself a
532 finding that deserves to be more clearly documented in the geologic CO₂ storage literature, but it may
533 also have large ramifications on what geologic formations are targeted for CO₂ storage because
534 monitoring costs are largely a function of plume size.
535

- 536 3. *Sites having higher geothermal gradients could have reduced sequestration costs unless brine treatment/disposal costs are*
537 *included, at which point sites with lower geothermal gradients may cost less.* The geothermal gradient also has
538 unexpected impacts on CO₂ injection and storage rates and costs. Our results show that an increasing
539 geothermal gradient generally reduces CO₂ storage costs because the well injection rate increases faster
540 than the reduction of stored CO₂ (owing to reduced density of CO₂). This would suggest, all things being
541 equal, that geologic formations having higher geothermal temperature gradients should be targeted for
542 geologic CO₂ storage before those with lower geothermal temperature gradients. However, this
543 relationship inverts when including the cost to extract and treat brine (Experiment #6; Figure 8).
544 Extracting brine might be needed to minimize induced seismicity risks while increasing the CO₂ storage
545 potential by freeing-up pore space. Treating brine not only adds an additional cost regardless of the
546 geothermal gradient, costs rise with an increasing gradient (as opposed to falling without brine treatment)
547 because each tonne of stored CO₂ displaces an increasingly larger volume of brine. This could have
548 substantial implications for choosing geologic reservoirs for CO₂ storage. Future studies could consider
549 treating the brine—particularly hot mineral-laden brines—as a resource that could help mitigate the cost
550 of CCS instead of adding to it. For example, it could be possible to chain sequestration, geothermal
551 energy (electricity or direct heat), mineral extraction, and water treatment technologies together to
552 produce an integrated system that uses the economic value of the brine (heat, minerals, and freshwater)
553 to offset sequestration costs or to increase system profitability.
554
- 555 4. *Allowing plumes to overlap has a significantly positive impact on increasing storage capacities but only a small impact on*
556 *reducing sequestration costs.* In other words, motivation for designing CO₂ storage sites with overlapping CO₂
557 plumes should be driven by the need to increase the amount of CO₂ stored at the site, and less by the
558 need to reduce the \$/tCO₂ cost. To our knowledge, this is the first study to investigate the effect of
559 letting CO₂ plumes overlap and the subsequent impact on storage rates and economics. Our results
560 suggest that the marginal increase in CO₂ storage capacity resulting from allowing CO₂ plumes to overlap
561 by 50% may be an order of magnitude larger than the resulting marginal decrease in cost, in part, because
562 the costs related to drilling and operation of the wells do not change if plumes overlap (Experiment #7;
563 Figure 10). That is, there is only a moderate reduction in cost by packing wells closer together even
564 though there is a significant increase in storage capacity. However, there could be other significant drivers
565 to place wells closer together such as more efficiently using the land surface to minimize environmental
566 impacts or to deal with fewer landowners or pore space owners. Further, the *SCO₂T* tool is a screening or
567 feasibility tool and does not take into account the pressure interaction between plumes and instead
568 assumes that pressure is being managed through brine extraction.

569 Overall, the seven experiments illustrate that there are complex interactions between geologic variables and
570 sequestration engineering and economics, and that this will almost certainly have a significant impact on
571 society reaching global sequestration goals. The experiments indicate that design and operational decisions—
572 maximum injection rates, brine treatment, and overlapping CO₂ plumes—can have a significant impact on
573 the feasibility of site-scale geologic sequestration and should be considered in future studies. Also, the results
574 show that, when trying to assess the suitability of potential sequestration sites, it is important to explore the

575 range of simulation outputs and not just to look at injection rates or storage area density or storage capacity
576 alone.

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585

7.0 FIGURES

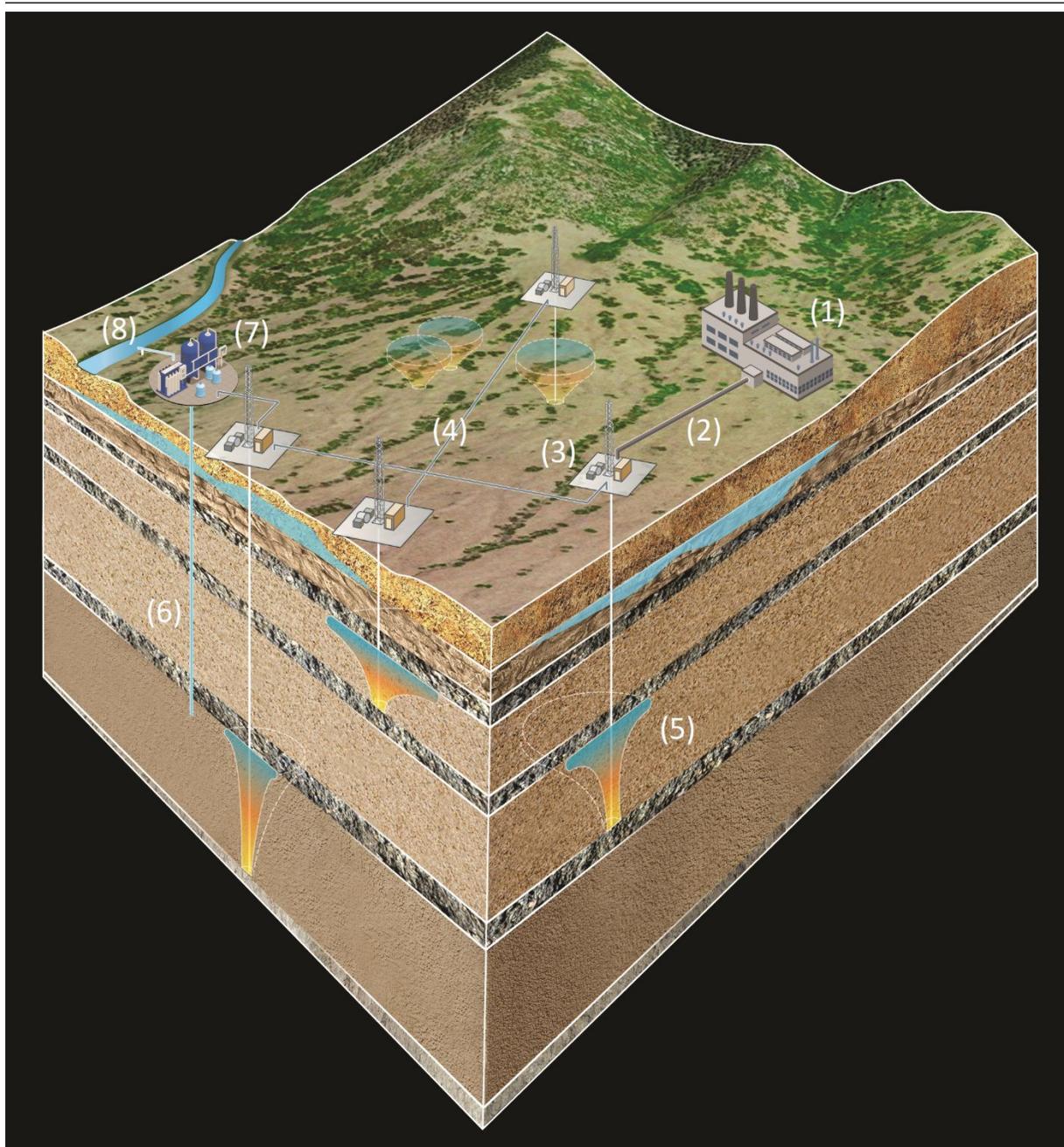


Figure 1: Overview of site-scale sequestration including (1) CO₂ sources such as a coal-fired power plant, (2) CO₂ pipeline network delivering CO₂ to the sequestration complex, (3) CO₂ injection well, (4) site-scale pipeline network distributing CO₂ between injection wells, (5) CO₂ plumes, (6) brine extraction, (7) brine treatment facility, and (8) end-use for treated water including disposal into surface water systems. The grey layers indicate caprock layers through which CO₂ cannot migrate, while the brown layers are indicative of CO₂ being stored in sandstone formations. The CO₂ plumes are colored for visual purposes only.

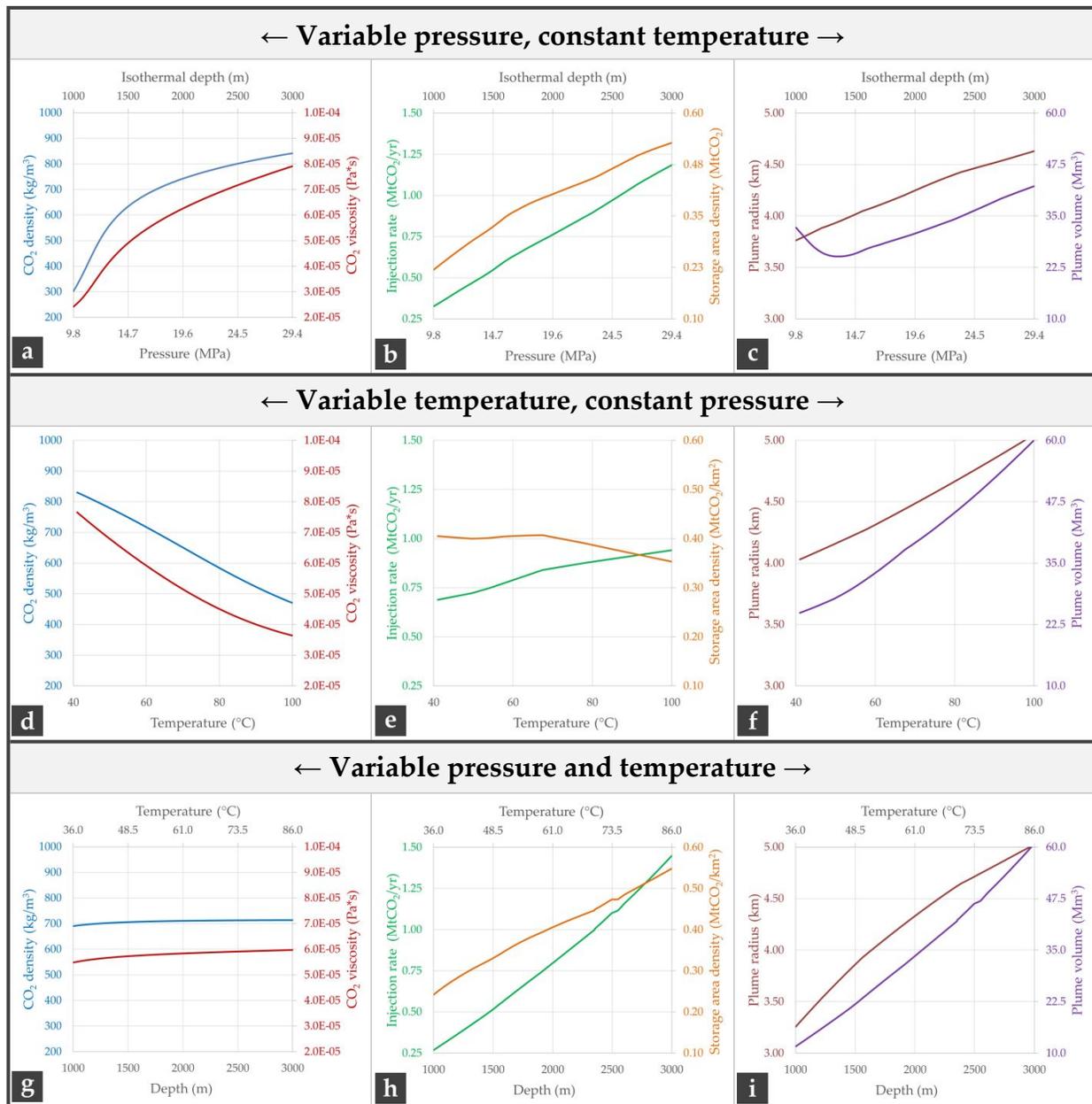


Figure 2: Impact of pressure-only (constant temperature; top row), temperature-only (constant pressure; middle row), and depth (pressure and temperature varying with depth; bottom row) on CO₂ density and viscosity (a, d, g), injection rate and storage area density (b, e, h), and plume radius and volume (c, f, i). Temperature is held constant at 66°C (associated with a formation depth of 2000 m) for the constant pressure runs. The y-axes in each chart have the same minimum and maximum values, whereas the x-axis always corresponds to pressure and temperature changes as depth increases from 1000 to 3000 m.

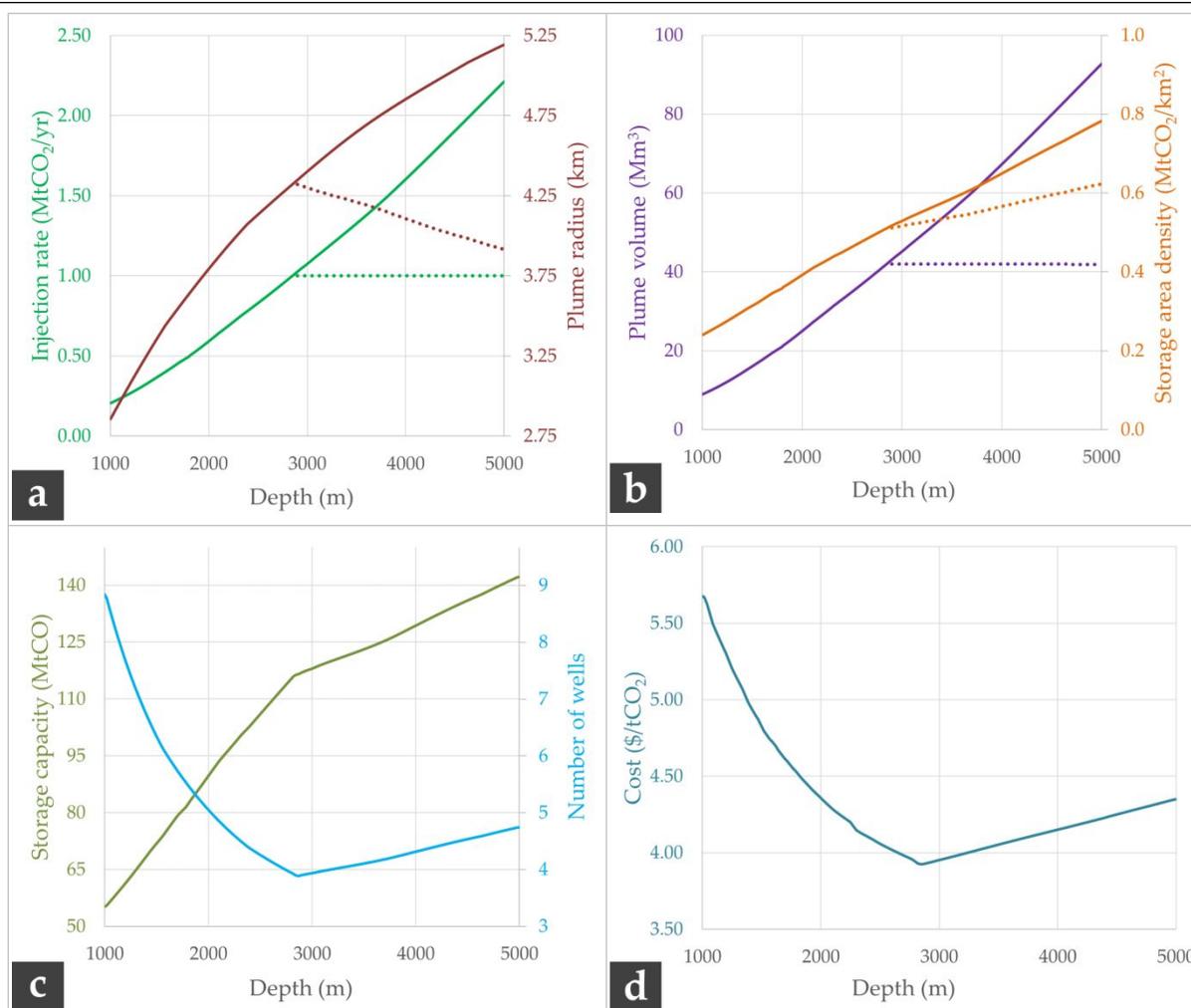


Figure 3: Impact of depth on: (a) injection rate and plume radius; (b) plume volume and storage area density; (c) reservoir storage capacity and number of wells used; and (d) CO₂ injection/storage costs. Depth ranges between 1000 and 5000 m while thickness (10 m), permeability (15 mD), porosity (0.2), and geothermal gradient (25°C/km) are held constant. Dotted lines indicate that the injection rate was capped at 1 MtCO₂/yr, illustrating divergences from the uncapped simulations (solid lines).

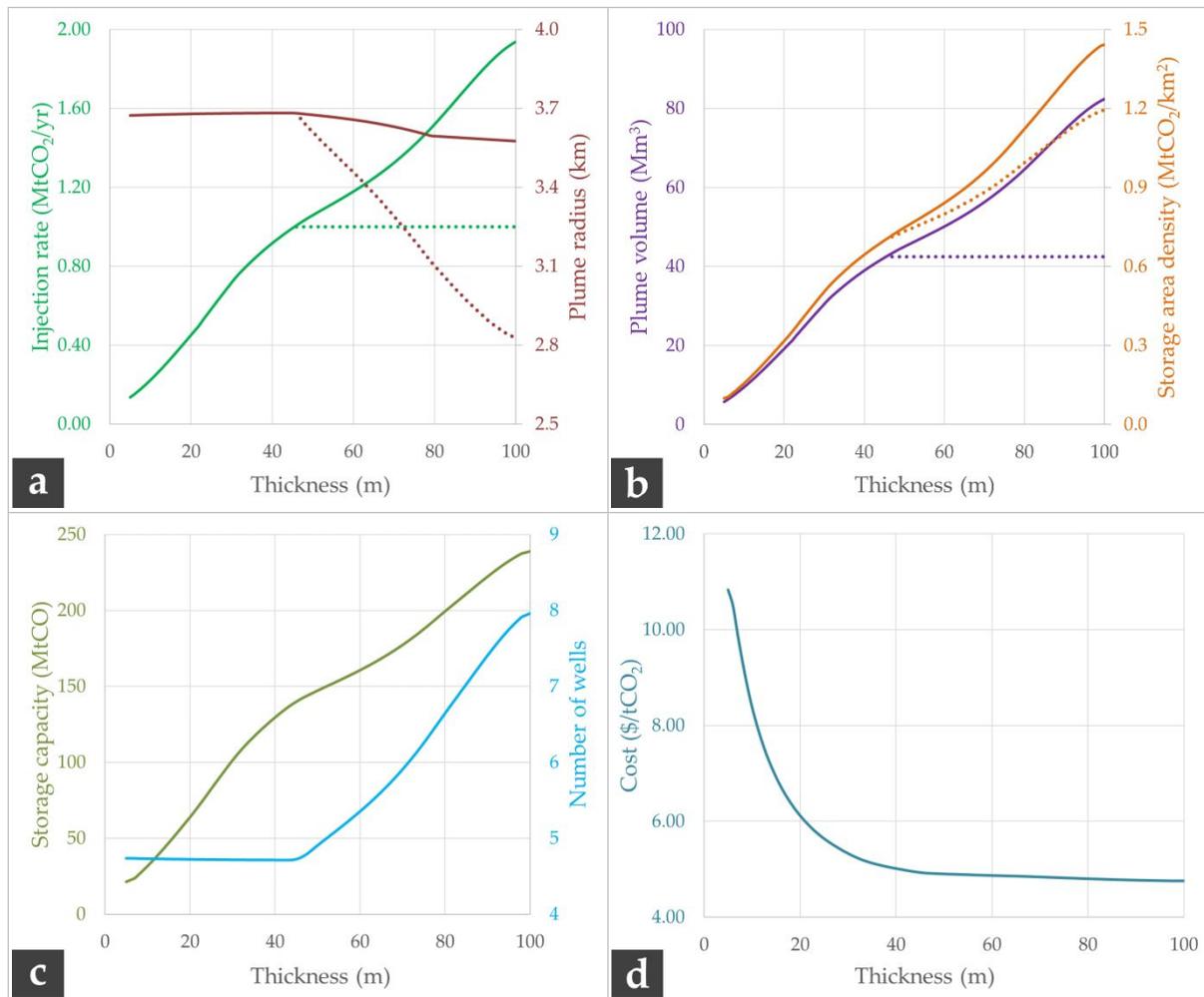


Figure 4: Impact of formation thickness on: (a) injection rate and plume radius; (b) plume volume and storage area density; (c) reservoir storage capacity and number of wells used; and (d) CO₂ injection/storage costs. Thickness ranges between 5 and 100 m, while depth (1500 m), permeability (18 mD), porosity (0.2), and geothermal gradient (25°C/km) are held constant.

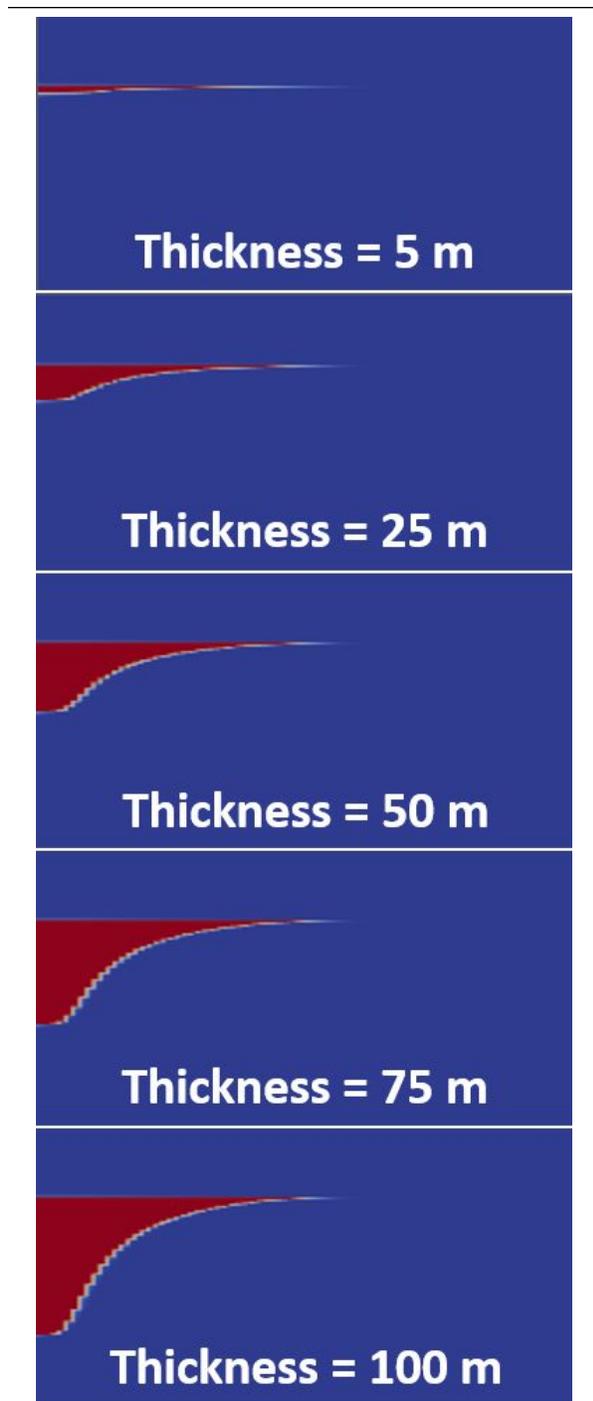


Figure 5: Impact of formation thickness on plume dimensions; full physics FEHM simulations for the reduced-order *SCO₂T* results in Figure 4. The vertical axis is exaggerated by five times. Red indicates the CO₂ plume.

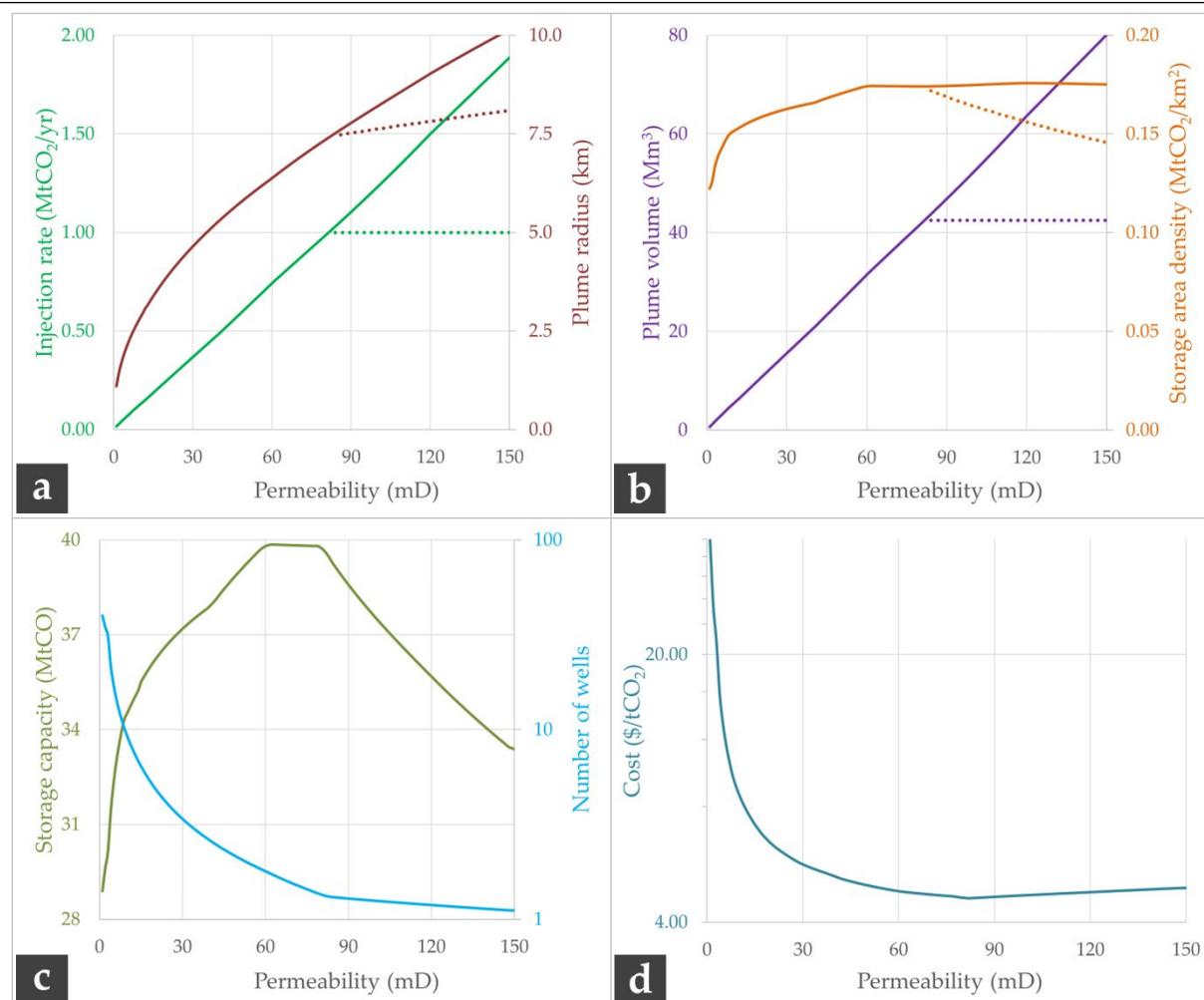


Figure 6: Impact of permeability on: (a) injection rate and plume radius; (b) plume volume and storage area density; (c) reservoir storage capacity and number of wells used; and (d) CO₂ injection/storage costs. Permeability ranges between 1 and 150 mD while depth (1500 m), thickness (10 m), porosity (0.2), and geothermal gradient (25°C/km) are held constant. Dotted lines indicate that the injection rate was capped at 1 MtCO₂/yr illustrating divergences from the uncapped simulations (solid lines).

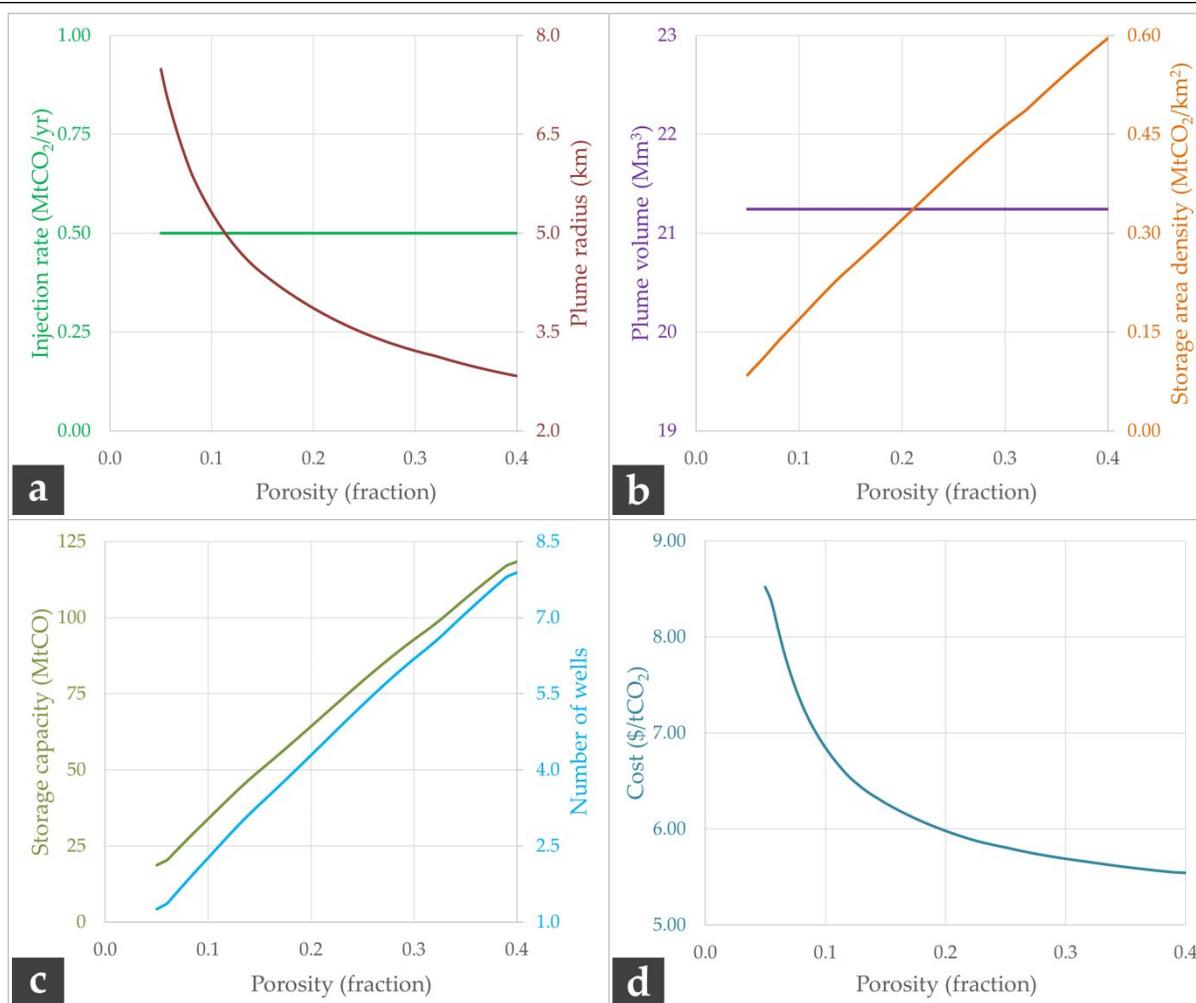


Figure 7: Impact of porosity on: (a) injection rate and plume radius; (b) plume volume and storage area density; (c) reservoir storage capacity and number of wells used; and (d) CO₂ injection/storage costs. Porosity ranges between 0.05 and 0.4 while depth (1500 m), thickness (10 m), permeability (20 mD), and geothermal gradient (25°C/km) are held constant.

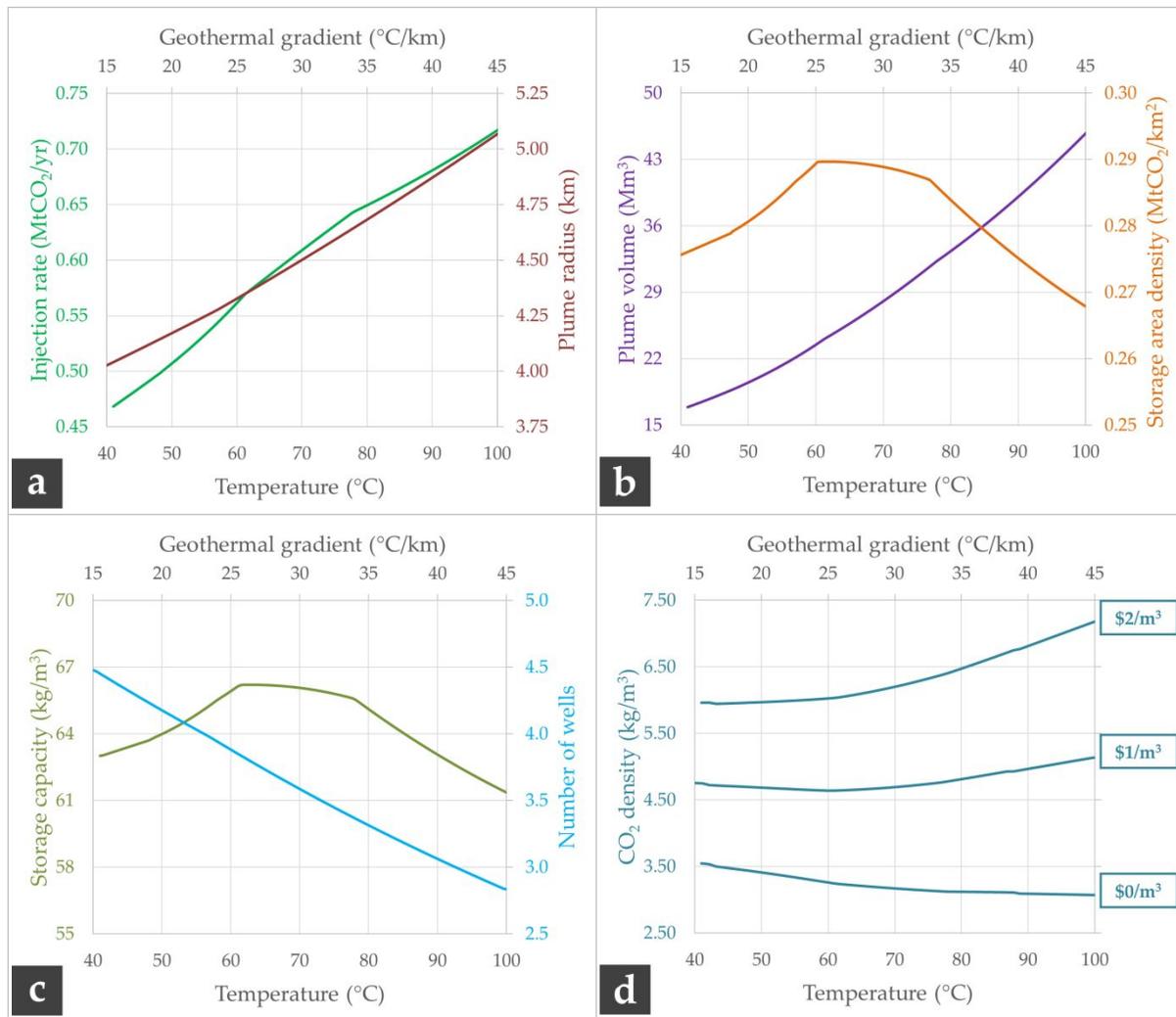


Figure 8: Impact of geothermal gradient on reservoir storage capacity and number of wells used (a) and CO₂ injection/storage costs (b). Geothermal gradient ranges between 15°C/km and 45°C/km while depth (2000 m), thickness (15 m), permeability (20 mD), and porosity (0.2) are held constant. CO₂ injection rate ranges between 0.5 and 0.75 MtCO₂/yr (i.e., does not reach the 1 MtCO₂/yr cap). For the injection/storage costs (b), three cost curves are shown with brine treatment/disposal costs ranging from \$0/tCO₂ to \$2/tCO₂.

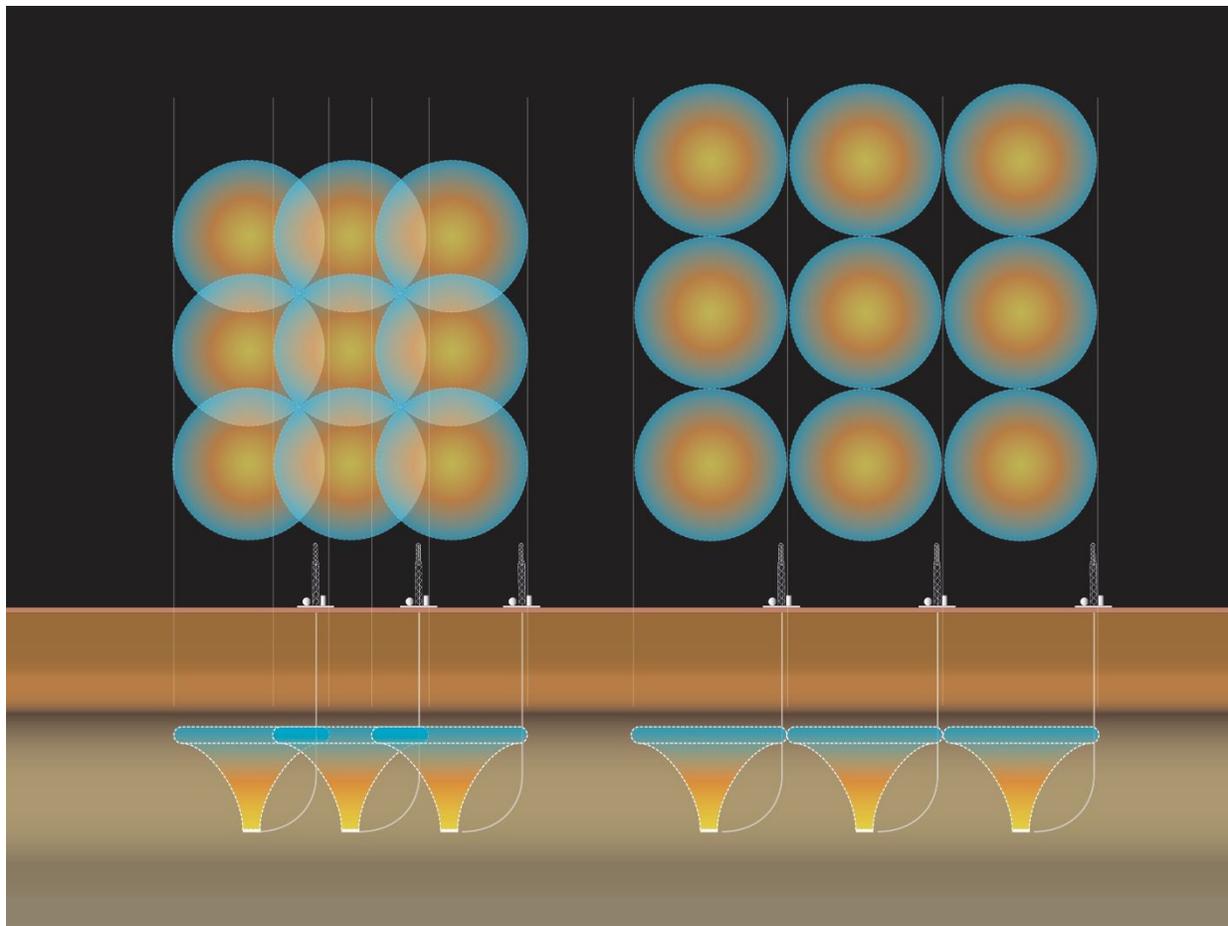


Figure 9: Conceptualization of overlapping plumes.

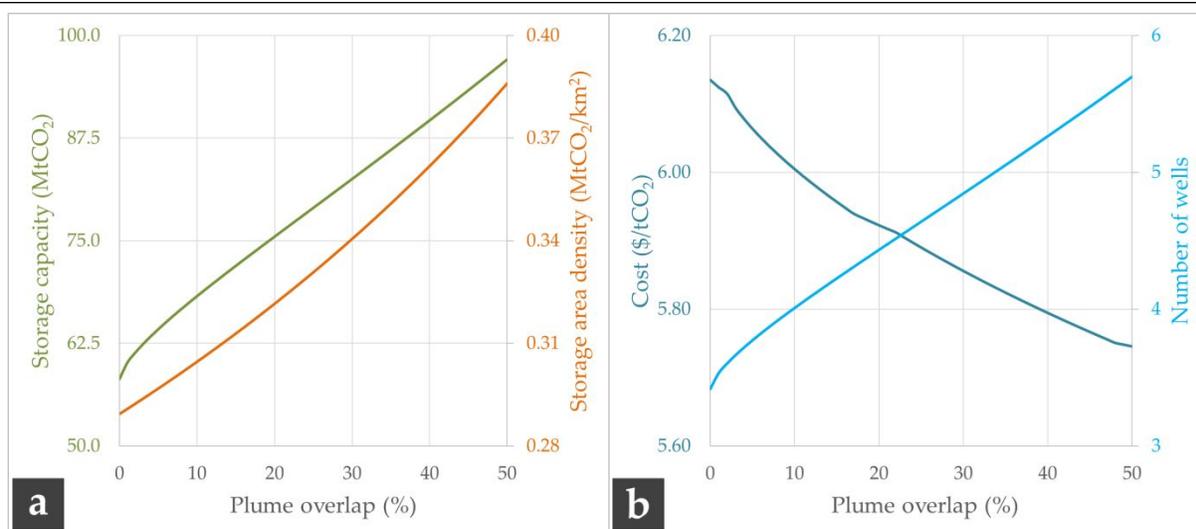


Figure 10: Impact of overlapping plumes (ranging from 0 to 50%) on storage capacity and storage area density (a) and cost and number of wells (b). While plume overlap ranges between 0 to 50%, depth (2000 m), thickness (15 m), permeability (20 mD), porosity (0.2), and geothermal gradient (25°C/km) are held constant. Wells have to be arranged in a square pattern (not the default hexagonal) in *SCO₂T* to allow the plume overlapping function.

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IMPACT

CO₂ capture and storage (CCS) is a critical technology for energy, the environment, and economic security. CCS is part of every clean energy pathway that limits global warming to 2°C and will change how we produce and consume energy. To have a meaningful impact, CCS will have to be deployed on a massive scale, potentially capturing and storing billions of tonnes of CO₂ each year around the globe. Identifying sites that can safely and cost-effectively store this amount of CO₂ relies on being able to quickly and accurately understand what large-scale CO₂ injection and storage. Our study focused on this critical research gap. We used the *SCO₂T* sequestration tool to explore the effect of key geologic characteristics (such as reservoir depth, thickness, permeability, porosity, and temperature) and operational decisions on the engineering and economics of carbon sequestration. We were able to show new and unintuitive effects that will be both important for fundamental environmental science research as well as understanding how we can meet climate and energy targets.