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Article

Simulation Analysis of CO₂-EOR Process and Feasibility of CO₂ Storage during EOR

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Abstract: In this study, oil production and retention were observed and compared in 72 reservoir simulation cases, after which an economic analysis for various CO₂ and oil prices was performed. Reservoir simulation cases comprise different combinations of water alternating gas (WAG) ratios, permeabilities, and well distances. These models were set at three different depths; thus different pressure and temperature conditions, to see the impact of miscibility on oil production and CO₂ sequestration. Those reservoir conditions affect oil production and CO₂ retention differently. The retention trend dependence on depth was not monotonic—optimal retention relative to the amount of injected CO₂ could be achieved at middle depths and mediocre permeability as well. Results reflecting different reservoir conditions and injection strategies are shown, and analysis including the utilization factor and the net present value was conducted to examine the feasibility of different scenarios. The analysis presented in this paper can serve as a guideline for multiparameter analysis and optimization of CO₂-enhanced oil recovery (EOR) with a WAG injection strategy.

Keywords: numerical simulation; EOR; CO₂ utilization and storage; mobility control



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1. Introduction

1.1. Introduction

Enhanced oil recovery (EOR) with CO₂ injection might be attractive because of the carbon dioxide retention in the reservoir [1], which provides a positive effect on the feasibility of CO₂ storage and storage capacity obligations regarding the European Union international agreements within the climate change domain (Kyoto protocol from the year 1997 and the Paris climate agreement from the year 2015). Carbon Capture Utilization and Storage (CCUS) comes into focus when the possibilities of CO₂ storage and reduction of storage cost are assessed. Although there are other utilization types, such as the utilization through beverage production or in agriculture, only the CO₂ enhanced oil recovery (CO₂-EOR) has been implemented at a commercial level on an industrial scale [2–4]. By injecting CO₂ above the miscibility pressure (or minimum miscibility pressure (MMP)), microscopic displacement efficiency is improved due to viscosity reduction, oil swelling, lower interfacial tension, and a change in the density of oil and brine [5]. The efficiency of CO₂ storage and CO₂ utilization factors are not easy to determine. Static approximations are easier to implement, which usually leads to the use of statistical distributions and stochastic models [6–8], sometimes accompanied by numerical simulation and parameter analysis [9,10]. Recently the increasing number of published works have been focused on more complex CO₂-EOR issues, such as injection rates of water and CO₂ (water alternating gas (WAG) ratios), permeability anisotropy, the effect of different simulation cell size, etc. [11,12]. Oil recovery, injection cost, and the amount of carbon dioxide permanently stored can be optimized by the application of methods, which include water alternating gas (WAG) injection. Simulation of the WAG process can help establish optimal relation between the stated parameters.

The injection scheme of a typical CO₂-EOR process can be classified according to:

1. CO₂ and oil miscibility
 - a. miscible
 - b. near-miscible
 - c. immiscible
2. injection type
 - a. continuous gas injection (CGI)
 - b. water alternating gas injection (WAG)
 - c. simultaneous water and gas injection (SWAG)

There are no guidelines for the analysis or selection of WAG ratios, well distance, permeability, and time of primary production parameter based on multi-case simulation study as an input. The main reason for the absence of such guidelines, and in general, the reason such analysis has not been performed, is the long run-time of a typical compositional reservoir model and/or high dependence on geological and fluid properties in the case of complex heterogeneous reservoir models.

When CO₂-EOR is performed, some of the CO₂ retained in the reservoir. Depending on policies, in some cases (for example, the Weyburn Project), it is considered CO₂ storage. For that reason, we use the word 'CO₂ retention' as a synonym for storage. This paper brings a novel approach by a number of variations of the simulation parameters in a conceptual model (i.e., reservoir model and reservoir settings that are similar to real cases but generalized for parameter sensitivity analysis).

CO₂ injection into a reservoir can be implemented in miscible and immiscible conditions, and the distinction between the two is defined by the minimum miscibility pressure. If the conditions are miscible, CO₂ increases the oil mobility as it is dissolved in oil between the injector and the producer, and if the conditions are immiscible, there is no CO₂ dissolution in oil, which means that CO₂ flows much faster than oil toward the producers, causing lower oil recovery and higher production of previously injected CO₂. The value of the minimum miscibility pressure depends mostly on oil composition and the reservoir temperature, and to determine the exact value of the minimum miscibility pressure, a detailed Pressure-Volume-Temperature (PVT) characterization of oil and the mixture of oil and CO₂ is necessary.

1.2. Literature Review

Klinkenberg and Baylé [13] studied pore size distribution and miscible and immiscible fluid injection and concluded that the pore distribution affects miscible and immiscible displacement differently.

Hall and Geffen [14] made a mathematical model for saturation pressure, using fluid flow velocity (of liquid and gas), and liquid phase fraction during the two-phase flow. Additionally, pure compounds (methane, propane, butane, etc.) were used in the analysis, and for these compounds, zones of gaseous state, two-phase region, and region of 100% liquid saturation were identified.

Lacey, Draper, and Binder Jr [15] studied the length of the mixed zone in cores of different diameters. It was concluded that the length of the mixed zone is proportional to the area through which the fluids flow, but this cannot be a rule at the reservoir level.

In the 1960s, different authors studied mixing mechanisms experimentally [16–18]. Benham, Dowden, and Kunzman [16] analyzed the miscibility of rich gases with reservoir fluid by using ternary diagrams with methane, C₂-C₄, and C₅₊. They developed a correlation for maximum concentration of methane in miscible conditions as a function of temperature, pressure, the molecular weight of C₅₊, and the molecular weight of C₂₊.

Other authors ([19,20]) used somewhat different components in the ternary diagram, but oil composition is, in general, divided into components of lower molecular weight, components of the medium, and components of high molecular weight. Such estimates disregard the reservoir heterogeneity (heterogeneity of permeability, pore structures, and fluid saturations), which are mentioned by Deffrenne et al. [21].

Peaceman and Rachford [22] proposed a numerical method for the calculation of the 2D miscible displacement. A normal permeability distribution was generated by a random number method, and the model gave results in line with measured experimental data.

The Buckley–Leverett theory was used as a starting point for numerous modifications. Koval [23] published the most known paper regarding the semi-analytical description of miscible processes, in which a model was given describing solubility as a function of pore volumes injected using the K-factor method for Buckley–Leverett equation extension. He compared the results obtained by the suggested mathematical model with published data and got satisfactory results for heterogeneous systems of the horizontal reservoir in which fluids of different viscosities flow at different velocities—viscous fingering.

Fitch and Griffith [24] tested alternating water injection to achieve more efficient oil displacement. Simon and Graue [25] used experimental data regarding solubility, swelling, and CO₂-oil system viscosity and gave a correlation for the prediction of these parameters as a function of oil viscosity and density. Different groups of authors developed more advanced mathematical, i.e., numerical simulation algorithms for CO₂ injection [26–29].

Rathmell, Stalkup, and Hassinger [30] analyzed the relationship between recovery, core length, and injection pressure on 13 m long cores, 5 cm in diameter, with 0.27 porosity and 1D permeability. They assumed that at certain parts of the core, immiscible displacement occurs even when the injection pressure was higher than the MMP. They found that different flow velocities occur due to different oil and CO₂ viscosities, which led to early CO₂ breakthrough and this effect was more pronounced in experiments done on shorter cores. They concluded that oil recovery depends on oil vaporization and higher fractions of swelling.

Teja and Sandler [31] used the equation of state in which they adjusted the binary interaction coefficients and applied adequate mixing rules to simulate the density of the system/mixture CO₂-oil, swelling factors, and CO₂ solubility in oil at a given temperature.

Wang [32] designed special equipment for visual detection of miscibility conditions, and he showed that miscible, near-miscible, and immiscible displacement could occur at the same time during the CO₂ injection. Among others, he also states that oil recovery cannot be the only criteria for MMP determination, and he proposed the determination of the optimal portion of CO₂ in a WAG process.

Sigmund, Kerr, and MacPherson [33] gave a simple correlation for relative permeability determination in a slim-tube simulation model

$$k_{ro} = \left(\frac{S_o - 0.15}{1 - 0.15} \right)^2 \quad (1)$$

$$k_{rg} = \left(\frac{S_g - 0.04}{1 - 0.19} \right)^2 \quad (2)$$

where k_{ro} is the relative permeability of the liquid phase, and k_{rg} is the relative permeability of the gas phase.

Li and Luo [34] used displacement on core samples and slim-tube experiments to determine a correlation for relative permeabilities determination, which represents input data needed for simulation. They tried to correlate Corey's exponents and the displacement pressure, but they concluded that relative permeability curves need to be adjusted by matching the simulation model with experimental data.

During WAG injection in water-wet rock, relative permeabilities depend on fluid saturation, saturation history, and the mobility will also depend on the interaction of viscosity, gravity, and capillary pressure [35]. Measurement of relative permeabilities during the three-phase displacement is usually not performed in a lab; therefore, it is common to measure relative permeabilities of a two-phase system, which is an acceptable input format for most commercial reservoir simulators.

Blunt [36] and Beygi et al. [37] reviewed most of so far suggested models for the determination of relative permeabilities of three-phase systems.

Jahangiri and Zhang [38] modified the second term of the optimization function (given by [39]) by introducing a mass of the stored CO₂ (concerning overall CO₂ mass storage capacity of the reservoir)

$$f = w_1 \frac{N_p}{OIP} + w_2 \frac{M_{CO_2}^S}{M_{CO_2}^T} \quad (3)$$

where w_1 and w_2 are weighted for oil recovery and CO₂ storage in the objective function (dimensionless), N_p is cumulative oil recovery (m³), OIP is oil in place (m³), $M_{CO_2}^S$ is mass of CO₂ stored (kg) $M_{CO_2}^T$ is the total storage capacity of reservoir (kg).

However, the overall CO₂ storage capacity of the reservoir is an uncertain parameter [40], and there is still no adequate optimization function for oil recovery and CO₂ sequestration.

WAG injection can increase oil mobility, increase the displacement efficiency and the oil recovery, but the usual problem of a WAG process is the reduced displacement efficiency due to water blockage of CO₂-oil contact [41], and that is the main reason why it is crucial to design the injection process correctly. Christensen, Stenby, and Skauge [41] gave an overview of 59 WAG projects. The expected recovery increase in some fields (South Swan, Slaughter Estate, Dollarhide, and Rangely Weber) is up to 20 %. Most of the WAG projects started in the tertiary phase of the exploitation. In other words, only recent WAG projects in the North Sea have started in the earlier exploitation phase. Eighty percent of projects are performed under miscible conditions, and the ratio of water and gas injection is mostly 1:1. Usual problems of a WAG process have been described, such as injectivity reduction, early water and gas breakthrough, corrosion, different temperatures of injected phases, hydrates formation, etc.

As a CO₂-EOR process can be classified as immiscible, near-miscible, and miscible, the same classification may be applied to the WAG process. The water is used to maintain the reservoir pressure above the MMP, and to prevent the early breakthrough of CO₂ to the production wells. Near miscible or immiscible WAG injection implies three-phase flow for which, no matter the longtime application of WAG processes in the world, there is still no complete understanding of changes in fluid composition which happen during such flow [41]. Extensive experimental and simulation research results have been published related to the mechanisms of displacement in the WAG process and water and gas injectivity [42–52]. Relative permeability models with different conditions of wettability for a three-phase system were developed along with hysteresis models necessary for the simulation of displacement in a WAG process [36,37,48,53–60].

Within the ESCOM (Evaluation System for CO₂ Mitigation) project [61], numerous numerical simulations of hypothetical reservoirs were performed to define a mathematical model that would help to estimate the additional recovery and amount of CO₂ retained in the reservoir. Different analyses identified the most important parameters, after which sensitivity analyses of these parameters were performed.

The amount of CO₂ retained in the reservoir as part of the CO₂-EOR project will become important in the EU if trading with European emission allowances (EUA) will be possible on the emission market, EU emissions trading system (EU ETS). Vulin et al. [62] investigate development of EU ETS, historical EUA price volatility, and trends correlated with price movements of natural gas, coal, and oil to forecast long-term EUA price probability using momentum strategy and geometric Brownian motion. Same as oil price, EUA price is influenced by policy, but some dependency on natural gas price and consumption was detected.

Sustainable development is based on finding the balance between industry growth and the environment, i.e., generating the lowest amount of damage while operating industrial activities efficiently.

To prove that CO₂-EOR represents a feasible, mature, and clean carbon capture utilization and storage (CCUS) option, it is crucial to single out the most economically favorable option that can be applied considering the parameters crucial for CO₂-EOR operations. Earnings of a CO₂-EOR storage project come from oil production and avoided CO₂. Avoided CO₂ refers to the amount of CO₂ retained in the reservoir during the EOR project, and this volume of CO₂ can be considered as allowance and therefore used for trading in the EU ETS. Price of EUA on EU ETS can be observed as possible additional income besides oil production and on the other side highest costs are related to transport and injection of CO₂ in CCUS projects.

Macintyre [63] examine different processes, instrumentation, material, and operating considerations connected to the design of CO₂ compression, dehydration, and injection and concluded that the additional cost of corrosion prevention measures for CO₂ injection is insignificant compared to the total implementation costs of the Joffre EOR project, which was the first miscible CO₂-EOR project in Canada. He provided a real p-h diagram for the Joffre EOR project that requires only three stages of CO₂ compression up to 130 bar.

McCollum and Ogden [64] gave some recommendations and examples of how to calculate compression and pumping power requirements emphasizing lower power requirements for pumping when the CO₂ liquefies, i.e., for pressure and temperature conditions above CO₂ critical values, including the Capital costs (CAPEX) estimation for pumping and compression phase of CO₂ transport and injection.

Habel [65] observed the difference between reciprocating and centrifugal compressors for onshore CO₂ CCUS applications with a recommendation for assessment of the required number of stages in case of centrifugal compressors that have many advantages over reciprocating compressors for flows over 12 kg/s and outlet pressure up to 250 bar.

Desai [66] made some general observations and challenges in the case of CO₂ compression, including steps that can help companies to estimate their CO₂ compressor requirements.

2. Materials and Methods

CO₂ injection as a part of a CO₂-EOR project, especially in the case of WAG injection, brings additional oil recovery (AR) accompanied by CO₂ production, so it is important to consider produced CO₂ quantities using retention, defined as

$$retention = CO_{2_{injected}} - CO_{2_{recycled}} \quad (4)$$

where CO_{2_{injected}} represents CO₂ injected through injection wells in Mt, and CO_{2_{recycled}} represents recycled CO₂ in Mt.

CO₂ produced at the production wells can be injected into the reservoir again after the separation process. This CO₂ represents recycled CO₂, which is injected together with the new CO₂ in the next injection cycle. Since the results are sensitive to multiple parameters,

and neither retention nor CO₂ recycled does not show all the details of the injection process, we defined a dimensionless variable storability as:

$$\text{storability} = \frac{\text{retention}}{\text{CO}_2 \text{ recycled}} \quad (5)$$

Optimization of the CO₂ injection in CO₂-EOR projects can be evaluated through CO₂ utilization factors (UF) since that factor is the ratio of utilized CO₂ and produced oil, i.e., in this paper, CO₂ UF was used and is previously defined by some authors ([67–69]) as

$$\text{UF} = \frac{(q_{\text{CO}_2 \text{inj}} - q_{\text{CO}_2 \text{prod}}) [\text{Mscf}]}{q_{o \text{prod}} [\text{stb}]} \quad (6)$$

where $q_{\text{CO}_2 \text{inj}} - q_{\text{CO}_2 \text{prod}}$ refers only to the new CO₂.

As for any other project, during the planning of an EOR project, it is common to develop an economical screen and perform discounted cash flow analysis, which is used as a standard method for deploying the time value of money in order to evaluate long-term developments, i.e., net present value (NPV) [70]

$$\text{NPV} = \sum_{t=0}^n \frac{R_t}{(1+i)^t} \quad (7)$$

where R_t is the net cash flow during period t , i is the discount rate, and t is the number of time periods.

Reservoir simulations are performed with Schlumberger's compositional numerical reservoir simulator E300, which uses the finite volume method to calculate the fluid flow.

The reservoir has a closed boundary, and the selected grid for all models has 29 cells in the x-direction (NX = 29), 29 cells in the y-direction (NY = 29), and nine cells in the z-direction (NZ = 9), wherein the dimensions of the cells in the x and y directions are 50 m both, while the dimension of the cells in the z-direction is 10 m.

The model is based on the Miocene sandstones in the Sava depression, where porosities range from 21.5% to 23.6%, and in our conceptual models, we used 23.5%.

The simulations for this study were divided into three different models differentiated by depth, which are 715, 1545, and 1845 m. Each of these three listed models was then broken down for cases of the greater and smaller distance between production and injection wells and three different permeabilities (5 mD, 50 mD, and 250 mD), as shown in Table 1.

Table 1. Parameter matrix for numerical simulation case-sensitivity analysis.

	Model 1	Model 2	Model 3
Depth (m)	715	1545	1845
WAG Ratio	1:1, 1:2, 2:1	1:1, 1:2, 2:1	1:1, 1:2, 2:1
Initial Datum Pressure (bar)	78	164	195
Reservoir Temperature (°C)	60	96.6	110
Permeability (mD)	5, 50, and 250	5, 50, and 250	5, 50, and 250
Distance between injection and production wells	Smaller and Greater	Smaller and Greater	Smaller and Greater

The differences between smaller and greater distance models per each production phase and model workflow are shown in Figures 1 and 2.

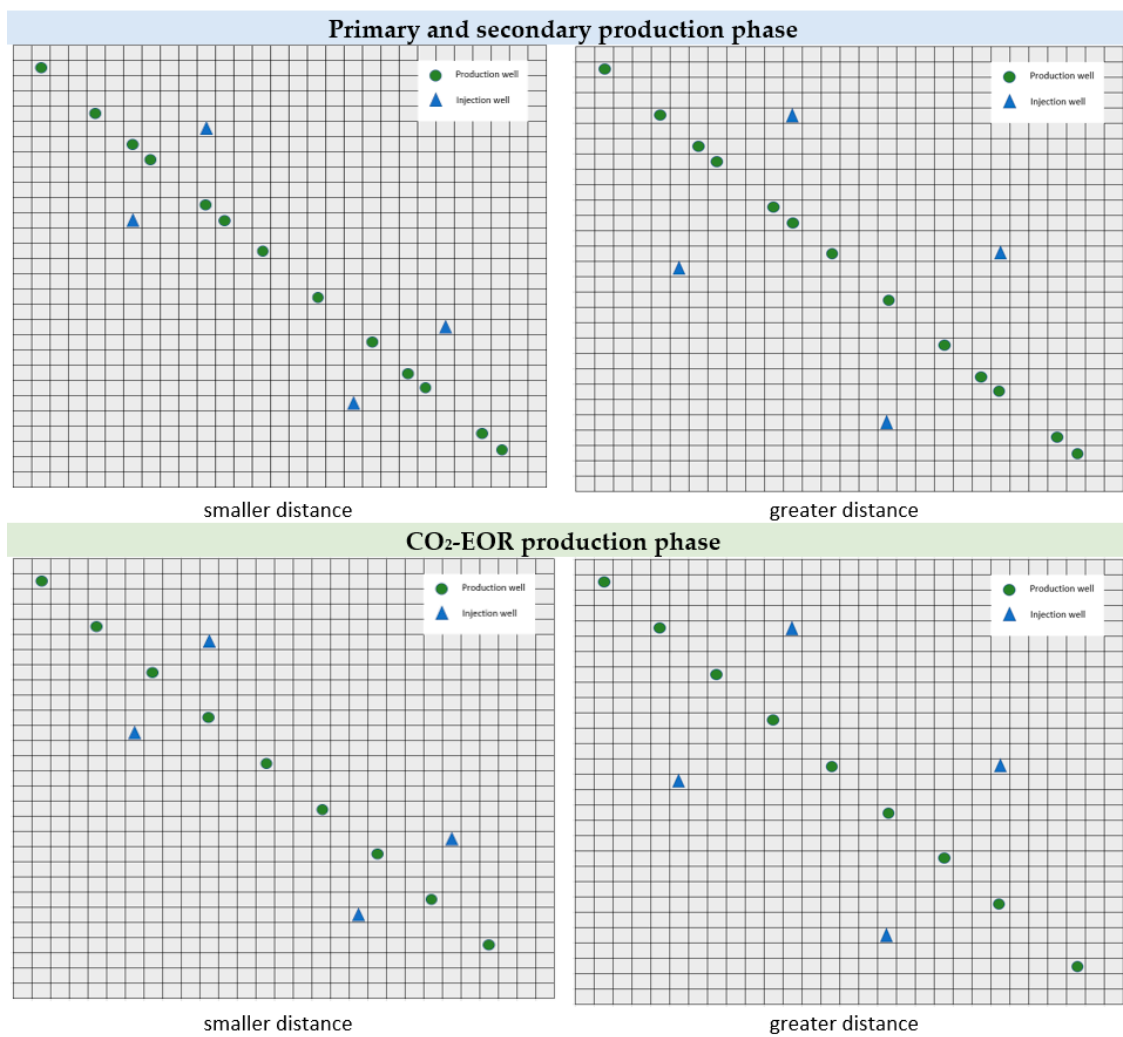


Figure 1. Models grid scheme. CO₂-EOR: CO₂-enhanced oil recovery.

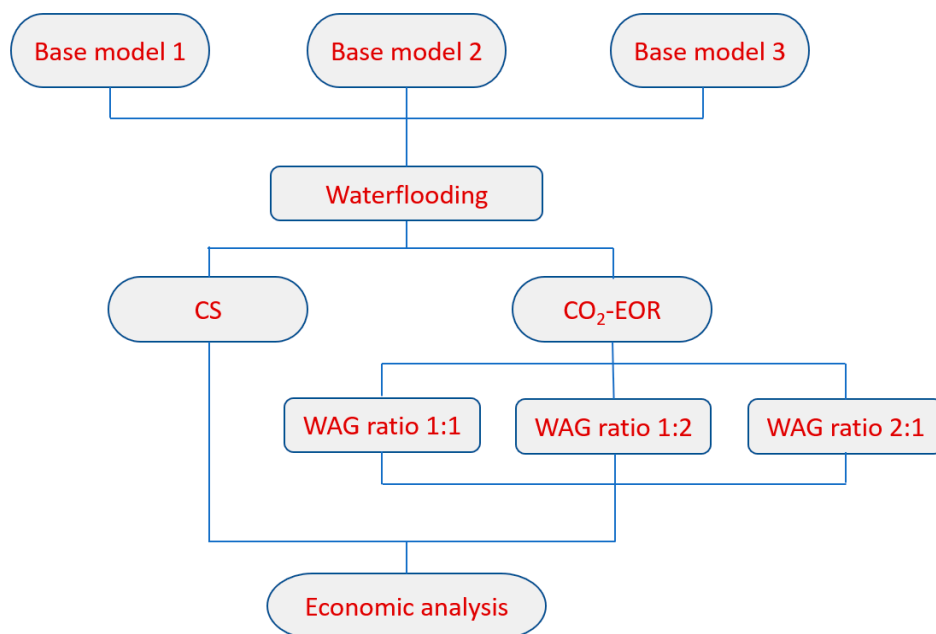


Figure 2. Model development workflow.

Base model fluid production and pressure are matched to total oil production. No problems were encountered in simulation cases in control of Newton and linear iterations, nor in time truncation and convergence controls (Table 2). The PVT model was fine-tuned, and matching of the EOS to the laboratory tests is presented in [71].

Table 2. Solver constraints and convergence controls used in the simulator.

Time truncation and convergence controls	
Target TTE (time truncation error):	0.2
Maximum non-linear convergence error:	0.001
Maximum material balance error:	0.000001
Target reduction in the square of the relative residual norm in the linear solver:	Improvement by a factor of 1×10^{-10} in the square of the residual
Maximum TTE (Time Truncation Error):	10
Target normalized solution change:	0.2 Fully Implicit
Maximum normalized solution change:	10 Fully Implicit
Maximum fugacity (phase equilibrium) error convergence criterion:	0.001
Control of Newton and linear iterations	
Maximum number of non-linear iterations in a timestep:	20 Fully Implicit
Minimum number of non-linear iterations in a timestep:	1
Maximum number of linear iterations:	40
Minimum number of linear iterations in a Newton iteration:	1
Maximum pressure change at last Newton iteration:	no-limit
Target maximum pressure change in a timestep:	100 atm

The fluid composition and characterization for Model 3 were the same as in already published work [71]. In that work, detailed data on a PVT study, including a slim-tube test, were presented, and the resulting EOS matched the CO₂ injection based on PVT laboratory data. PVT properties from that work were entered to the PVTp (Petroleum Experts software for PVT analysis), after which oil composition from the DLE (Differential Liberation) experiment at the pressure and temperature that corresponds to the initial pressure of each model was taken as the input composition for other two depths models (Table 3). Since the saturation pressure was 137 bar, the initial oil composition for Model 2 and Model 3 were the same. Model 3 corresponded to the oil reservoir in Sava Depression in terms of ranges of dynamic properties (mainly permeability and fluid properties).

Table 3. Oil composition for models.

Component	Model 1	Model 2 and Model 3
	mol%	mol%
N ₂	0.04	0.09
CO ₂	0.36	0.46
C1	20.30	33.25
C2	3.66	3.92
C3	3.31	3.11
NC4	3.23	2.83
NC5	3.32	2.81
C6	3.36	2.78
C7::13	8.80	7.24
C14::19	16.72	13.60
C20::25	17.64	14.29
C26::32	12.86	10.41
C33::C46	6.41	5.20

Based on compositions from DLE, MMPs were determined for each base case, i.e., slim tube experiment was simulated in PVTp to obtain oil recovery for different injection pressures after 1.2 pore volumes of CO₂ injected. MMP was defined as the intersection of two regression lines on a graph that relates the realized oil recovery and injection pressures. MMP for Model 2 remained similar as in Model 3, and in the case of Model 1, it was significantly lower (155 bar, Figure 3).

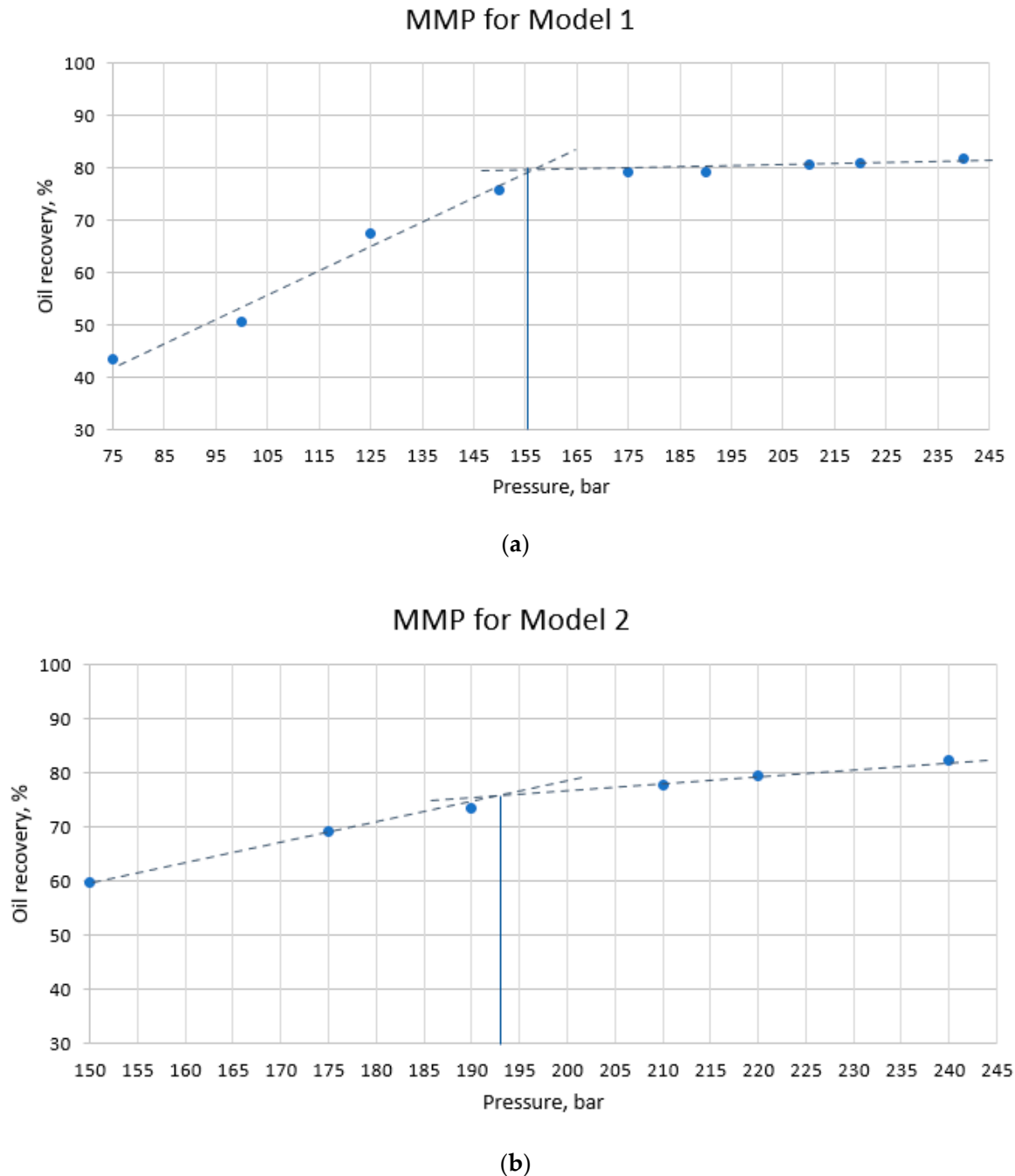


Figure 3. Minimum miscibility pressure for (a) Model 1 and (b) Model 2.

Three base cases (at three depths with respective initial reservoir pressure and then each with three different observation permeabilities) were simulated. Then cases were restarted after 10 years of primary production to simulate the secondary production phase for 48 years. After that, simulation cases were restarted again to simulate CO₂-EOR that lasts fifteen years.

Maximum injection pressure for water and CO₂ during the CO₂ EOR phase was set to 30% higher than the initial reservoir pressure to simulate immiscible, near miscible and miscible conditions for used oil properties.

Different input values for prices of oil, price of EUA, and discount rates were considered to cover different earnings scenarios from more pessimistic to more optimistic scenarios (Table 4).

Table 4. Input values for oil, carbon prices, and discount rate (r).

Parameter	Price	Price	Price	Price
Oil		25\$/bbl	40\$/bbl	50\$/bbl
CO ₂	10€/t	25€/t	40€/t	55€/t
r		8%	10%	12%

Capital costs (CAPEX) from the reviewed literature varied too significantly, and CAPEX was strongly site-specific. CAPEX for all the models was the same and taken from [72], since the models in this paper were similar to Ivanić field. The royalties, CAPEXs, operating expenses as an OPEX multiplier (i.e., percentage of the produced oil value, representing processing and transportation costs to the mid-stream), the cost of water, and the CO₂ injection used for economic evaluation are presented in Table 5.

Table 5. Input values for costs of development and implementation of EOR.

Parameter	Value
CAPEX	29 million €
OPEX, percentage of produced oil value	5%
Injection of water	1 €/t
Electricity price	20 and 40 €/MWh
Royalty, percentage of produced oil value	12%

CO₂ injection costs in this paper were calculated according to [64], implying five stages of compression, i.e., CO₂ injection costs are not fixed but depend on the injected amounts of CO₂ and the injection pressure as a base for an estimation of the energy required for compression and charges for that energy.

3. Results

Additional recovery and retention were opposed parameters, i.e., maximum retention does not always bring maximum additional recovery (Figure 4). In the first years of CO₂-EOR, retention to additional recovery ratio was higher, and with time, this ratio decreased.

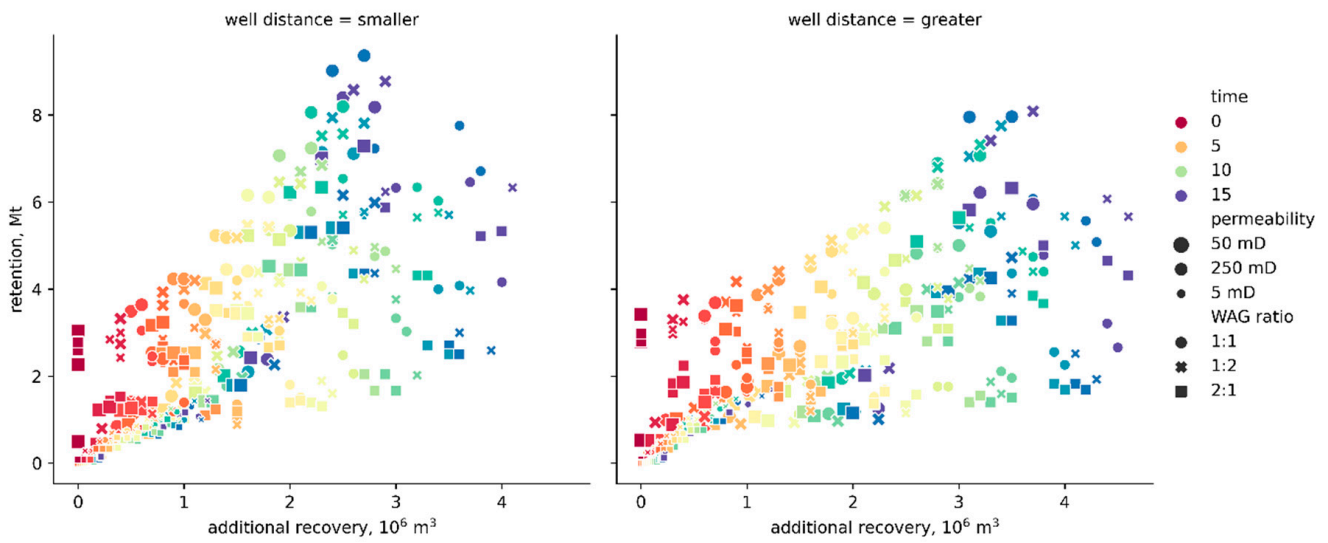


Figure 4. Retention versus additional recovery.

The results show that if additional recovery was observed in relation to the injected pore volumes (PVs) of CO₂ after 15 years of CO₂-EOR, it was obvious that more PV could be injected in cases with a smaller distance between production and injection wells, which finally, allows more CO₂ to be retained in a reservoir (Figure 5).

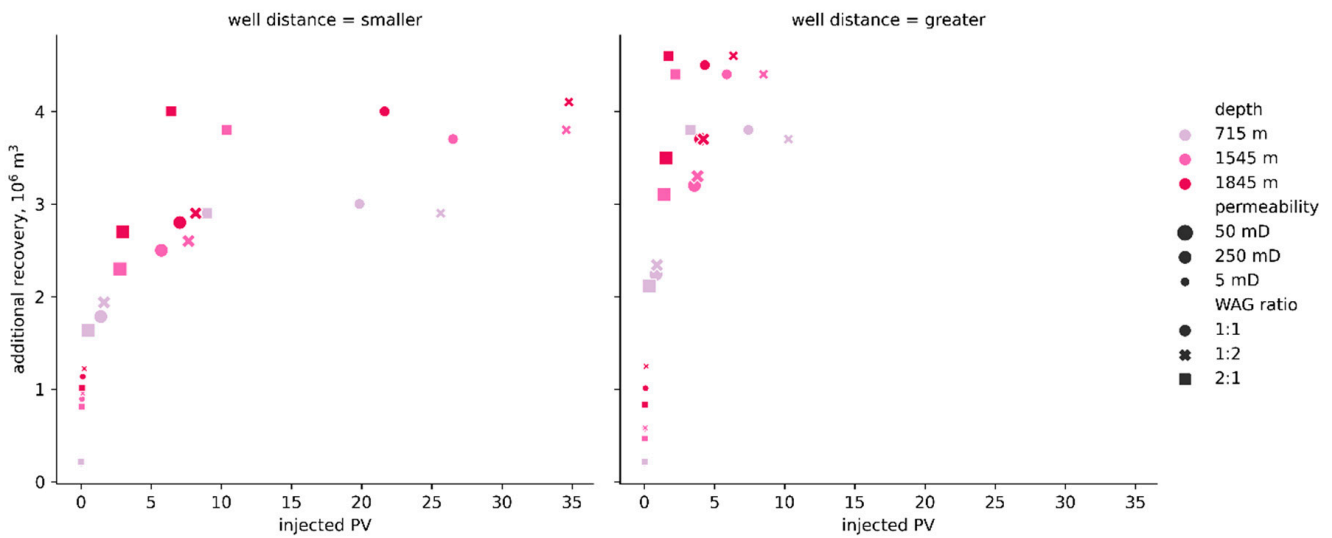


Figure 5. Additional recovery versus injected pore volumes (PVs) after 15 years of EOR.

Retention was also highest for the cases with smaller well distances and a WAG ratio of 1:2 since much more CO₂ could be injected in the case of smaller distances and a WAG ratio of 1:2 was the ratio in which CO₂ was injected for the longest period (Figure 6).

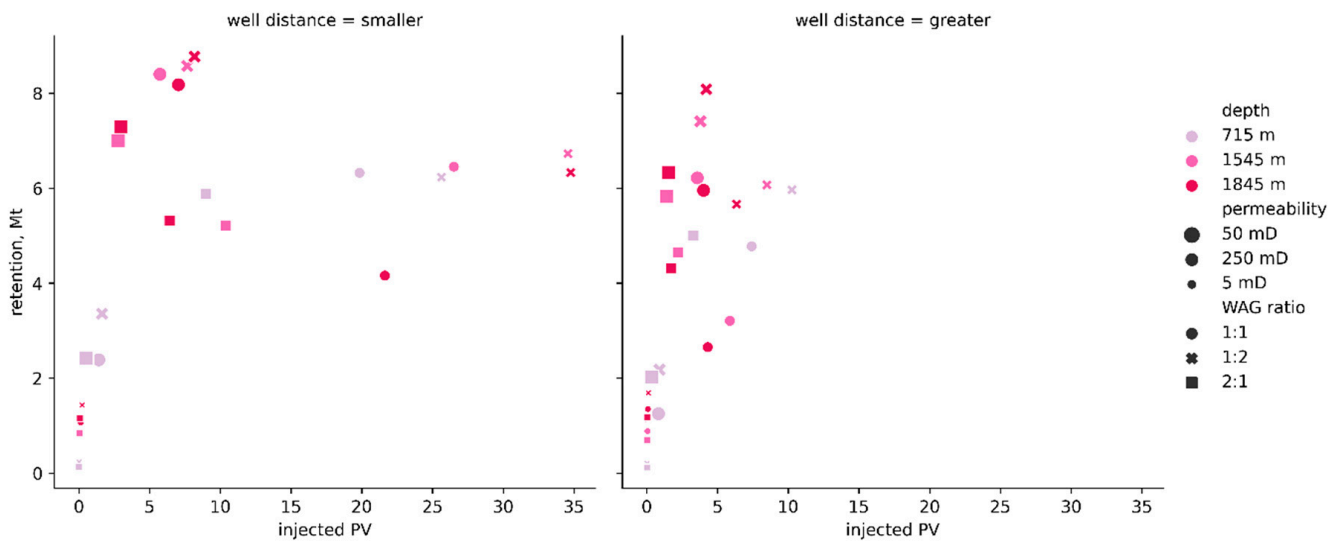


Figure 6. Retention versus injected pore volumes after 15 years of EOR.

The results showed that if retention was observed relative to the new CO₂, then the maximum retention was achieved at the greatest depths (Figure 7).

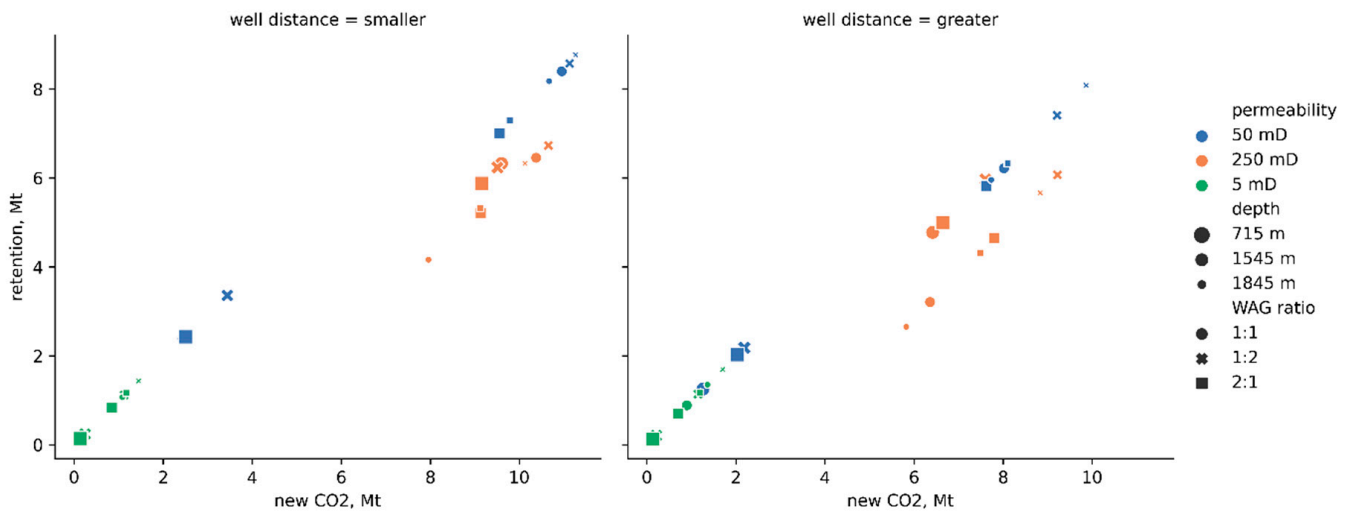


Figure 7. Retention versus new CO₂ after 15 years of EOR.

Higher storability was achieved at a greater well distance with less injected pore volumes, and the optimum WAG ratio was 2:1 (Figure 8).

Storability was inversely proportional to permeability (highest storability in the case of lowest permeability), and the greatest storability was achieved for the smallest amounts of new CO₂ brought to the system (Figure 9).

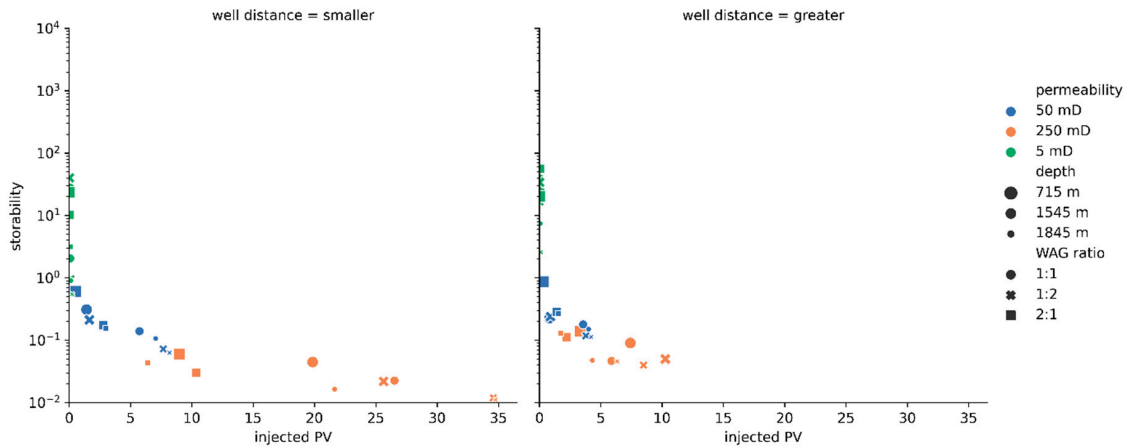


Figure 8. Storability versus injected pore volumes after 15 years of EOR.

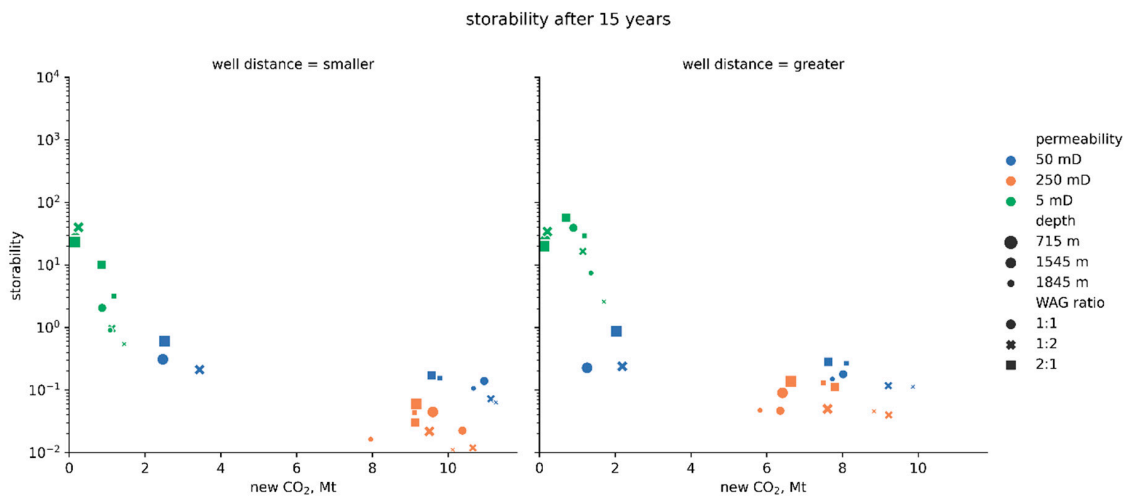


Figure 9. Storability versus new CO₂ after 15 years of EOR.

It should be noted that storage capacities for pure storage without EOR could not be achieved at the lowest permeabilities because of too high pressure (low injectivity) near the injection wells. The reason for that is that production wells act as some kind of "pressure release" wells during CO₂-EOR.

There was a decreasing trend of storability with PV injected (Figure 8), which can be described by power-law functions (coefficients given in Table 6). Moreover, if permeability was included in the correlation, a decreasing trend could be described by the power-law function as well:

$$storability = (c_1 \cdot k^{c_2} + c_3) \times (PV_{inj})^{c_4 \cdot k^{c_5} + c_6} \tag{8}$$

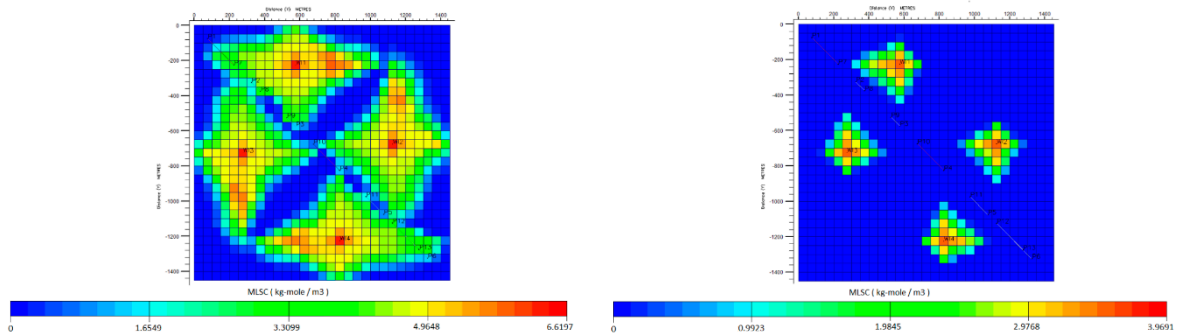
Table 6. Coefficients for storability correlation for the smaller and greater distances given in Figure 8.

	<i>c</i> ₁	<i>c</i> ₂	<i>c</i> ₃	<i>c</i> ₄	<i>c</i> ₅	<i>c</i> ₆
greater well distance	12.45440	−1.13065	0.16019	0.35021	0.03930	−1.06107
smaller well distance	0.842453	−0.24244	0.029734	7.927405	0.02378	−9.83669

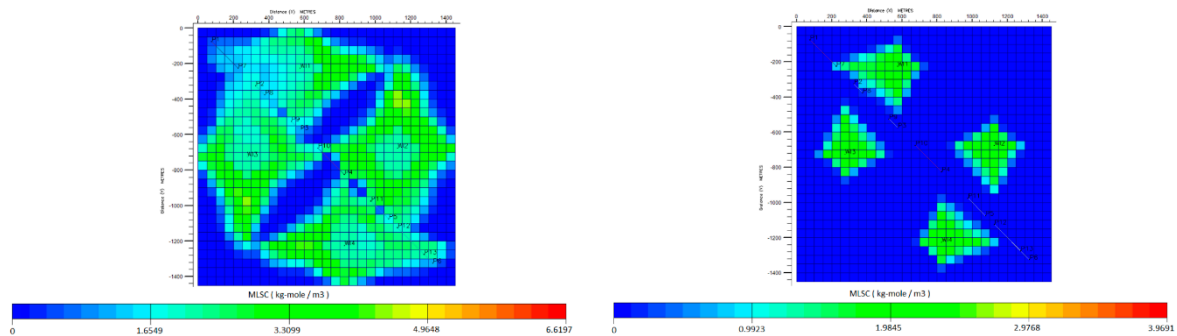
It should be noted that such a general correlation should be derived for each new reservoir setting (depth, well pattern, and oil composition) and each set of production parameters (WAG ratio, use of surfactants, and polymers).

Figure 10 shows two extreme cases of WAG injection. Left (Model 3) shows the CO₂ molar density change under fully miscible conditions (greater depth), while on the right side, the same is shown for the case where the CO₂ is injected under immiscible conditions (shallow reservoir). As CO₂ can appear both in a gaseous and liquid state (supercritical is treated as a liquid in Eclipse), molar density is the best way to visualize the change of composition in space. Favorable cases (like that on the left, Figure 10) will result in greater areal sweep efficiency and consequently in higher CO₂ retention (i.e., CO₂ stored).

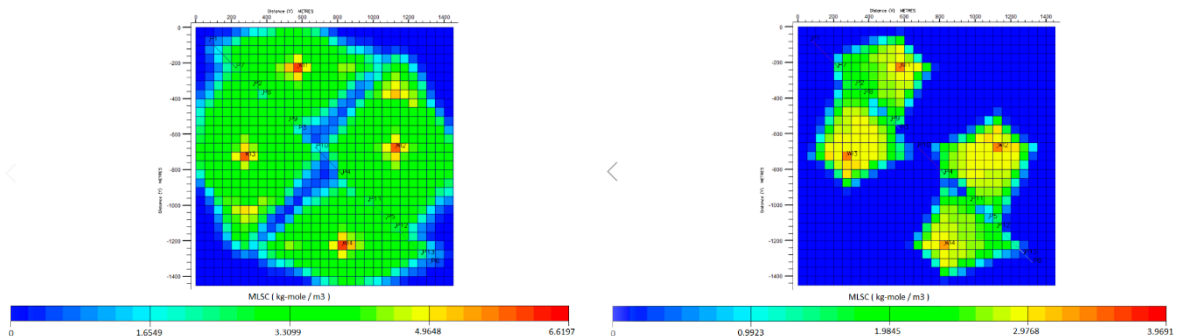
Model 3—Greater distance—Permeability 50 Model 1—Greater distance—Permeability 50
First year of EOR (CO₂ injection)



Second year of EOR (Water injection)



Seventh year of EOR (CO₂ injection)



Eighth year of EOR (Water injection)

Figure 10. Cont.

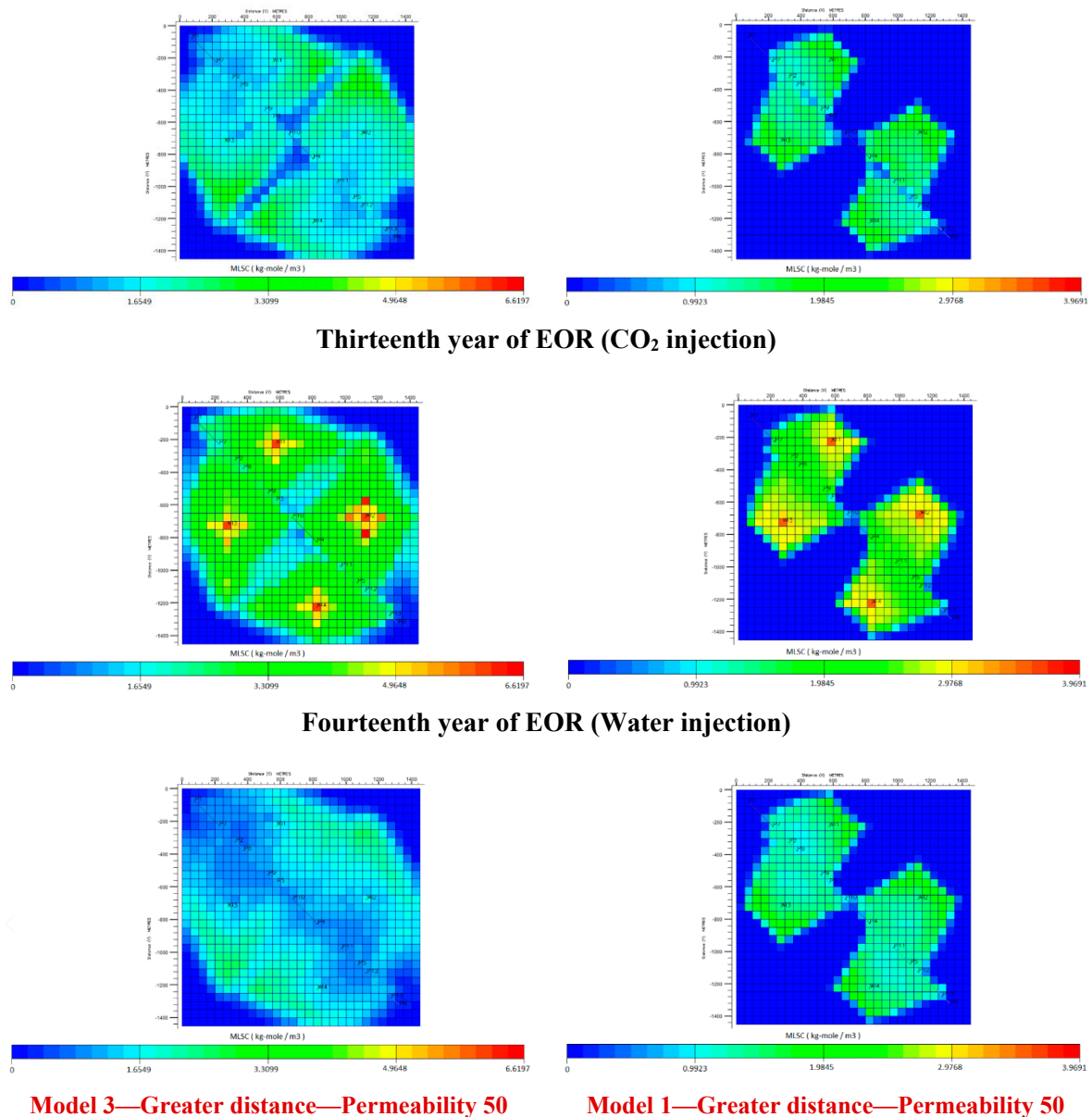


Figure 10. CO₂ molar density change during WAG injection with a WAG ratio of 1:1.

The utilization factor was significantly higher for cases with smaller well distances and was achieved for approximately the same injected PV's as in the case of greater distance (Figure 11). The highest UF was achieved at some mediocre values of injected PV, mediocre permeabilities ($k = 50$ mD), and medium depths (1545 m). The smallest UF's were achieved for lower depths and smallest permeability.

All previously presented parameters and their relationships should be presented concerning the economic value. The parameter that included both values added in terms of oil recovery and value-added in terms of CO₂ permanently retained in the reservoir was UF.

The realistic scenario is a scenario that implies current oil and EUA prices with the highest discount rate ($r = 12\%$). The highest discount rate is selected because of the risk connected with oil and EUA price movements, which are hard to predict since those prices are usually influenced by the world's political and economic situation. Also, electricity price varies differently throughout hours of the same day, with a maximum of 40 €/MWh and a minimum of 20 €/MWh, and that was the reason for calculating NPV for those two different scenarios of electricity price, i.e., costs of injection (Figures 12 and 13).

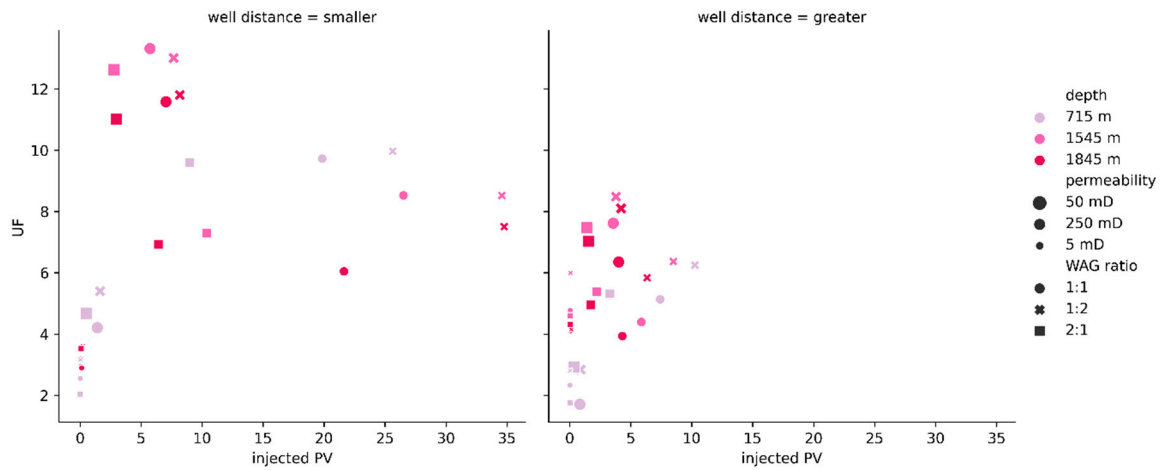


Figure 11. Utilization factor (UF) versus injected PV after 15 years of EOR.

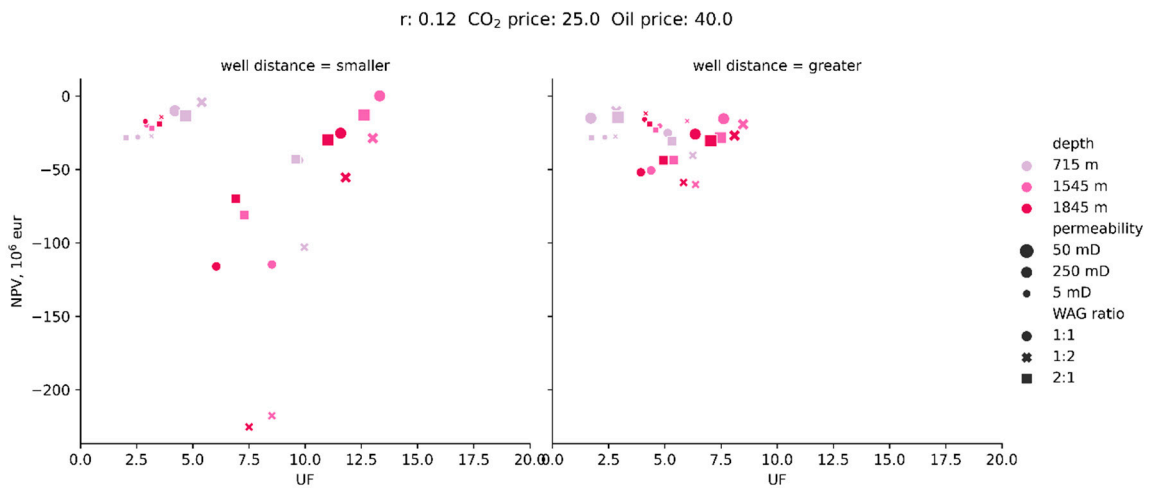


Figure 12. Net present value (NPV) versus UF for the realistic scenario at the end of EOR with 40 €/MWh electricity price.

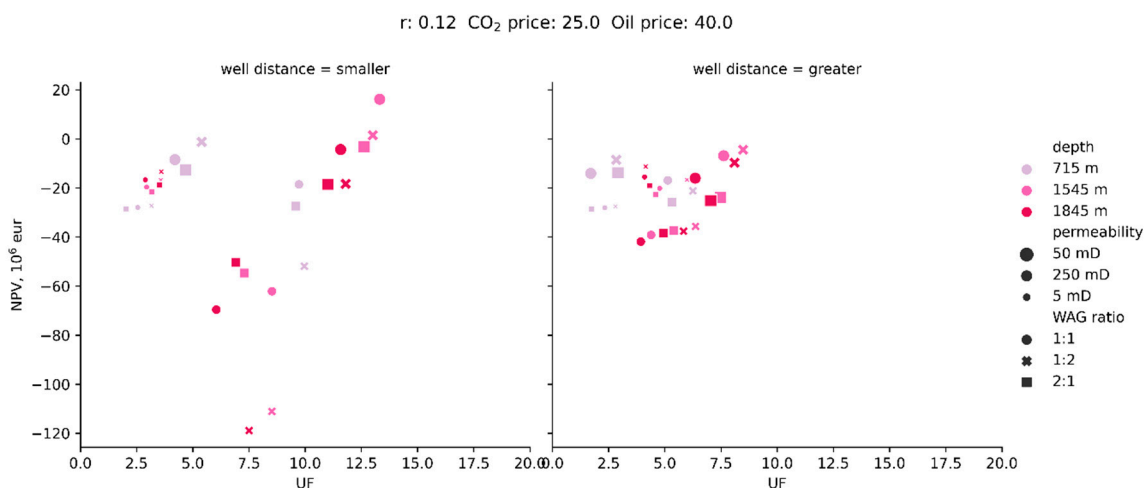


Figure 13. NPV versus UF for the realistic scenario at the end of EOR with 20 €/MWh electricity price.

The most optimistic case represents the highest prices of oil and carbon and the lowest discount rate (Figures 14 and 15). On the other hand, the most pessimistic scenario had the lowest prices for carbon and oil and the highest discount rate (Figures 16 and 17).

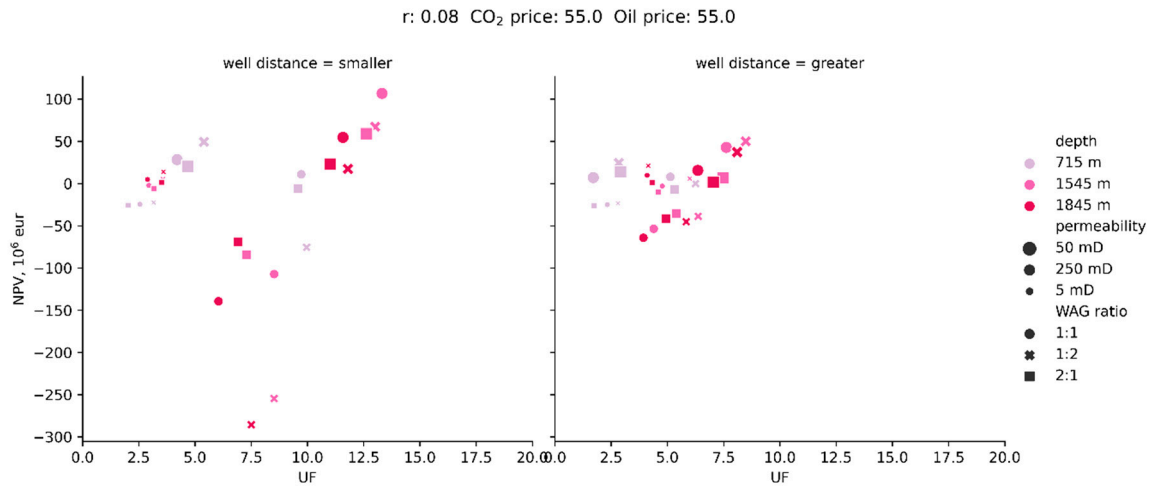


Figure 14. NPV versus UF for the optimistic scenario at the end of EOR with 40 €/MWh electricity price.

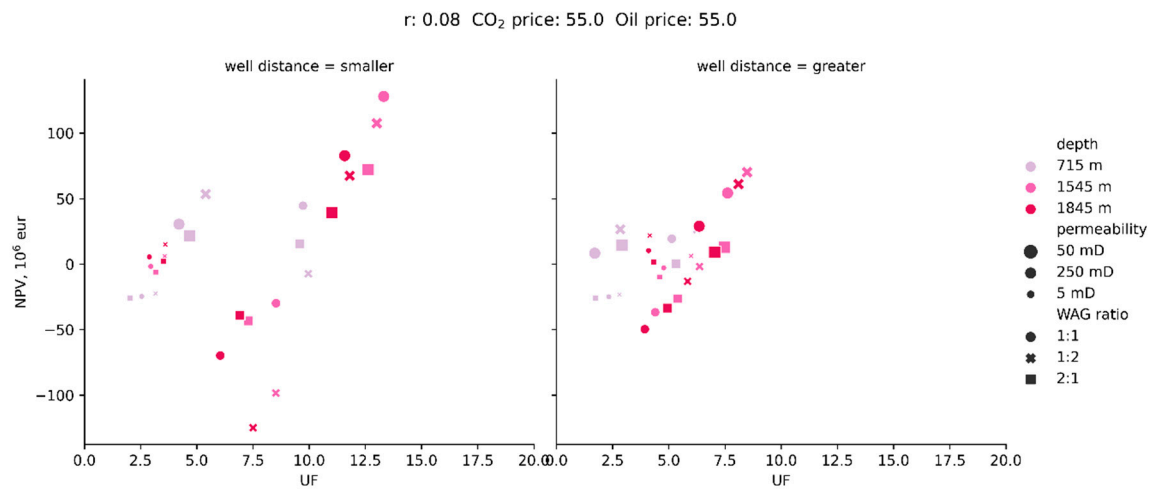


Figure 15. NPV versus UF for the optimistic scenario at the end of EOR with 20 €/MWh electricity price.

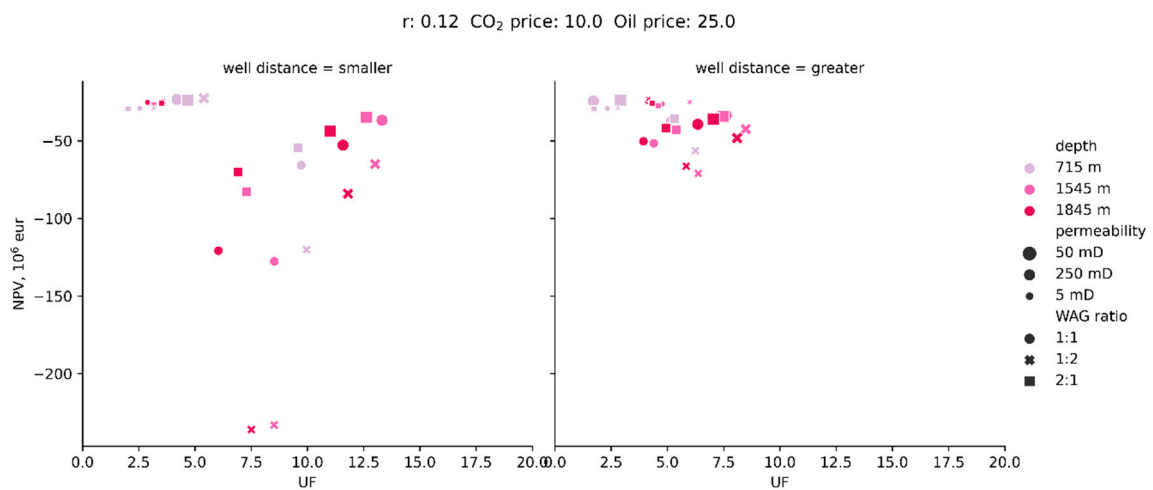


Figure 16. NPV versus UF for the pessimistic scenario at the end of EOR with 40 €/MWh electricity price.

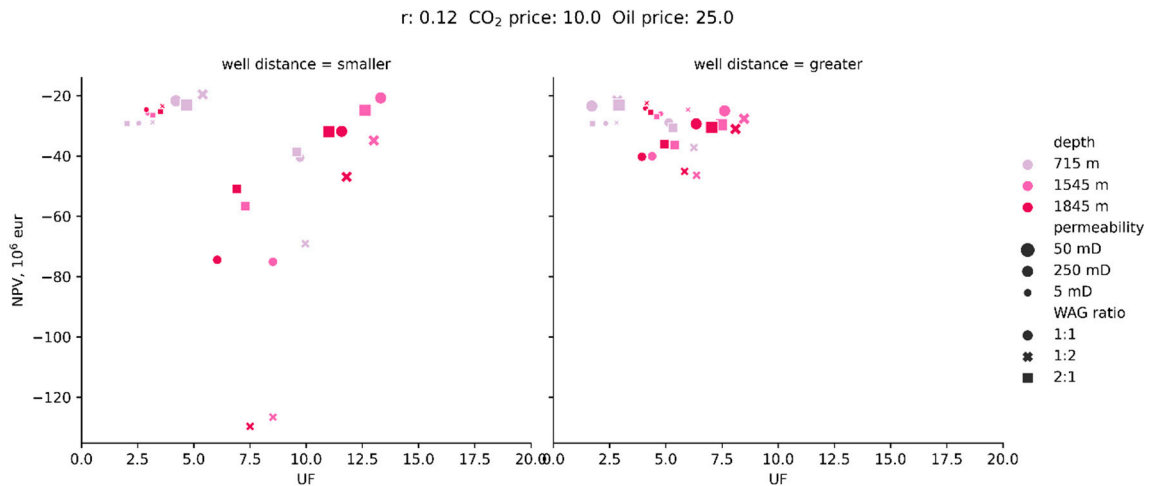


Figure 17. NPV versus UF for the pessimistic scenario at the end of EOR with 20 €/MWh electricity price.

For optimistic scenarios, both electricity prices scenarios resulted in positive NPV value after 15 years of production (for electricity price 40 €/MWh maximum value is 106.8 million € and in case of 20 €/MWh maximum value was 128.1 million €). In a pessimistic scenario, it is not possible to achieve a positive NPV no matter the electricity price. Based on more calculated price scenarios (totaling 3888 cases of NPV observation after 15 years of CO₂-EOR, Figure 18), more detailed insight into the influence of CO₂ and oil price could be obtained.

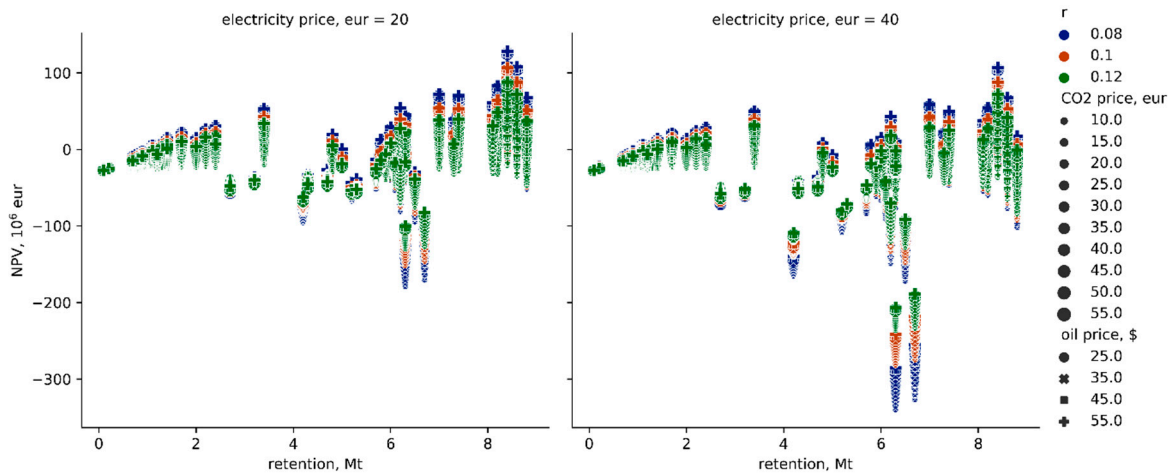


Figure 18. NPV versus retention after 15 years of CO₂ EOR for different discount rates, oil and CO₂ prices, and two different electricity prices (compression costs).

NPVs in time, for realistic and optimistic scenarios and for both electricity prices are shown in Appendix A (Figures A1–A4). NPV in the first-time step was negative due to capital investment, but over time it moved towards higher/more positive values, depending on the observed scenarios.

The impact of the retention on NPV in the case of higher and lower electricity prices was not significant given that the impact of electricity prices did not relate to the retained CO₂ but to the cost of injected CO₂.

While high retention was generally connected with higher NPV, some optimal cases resulted in the highest NPVs, but not the highest additional recovery (Figure 19).

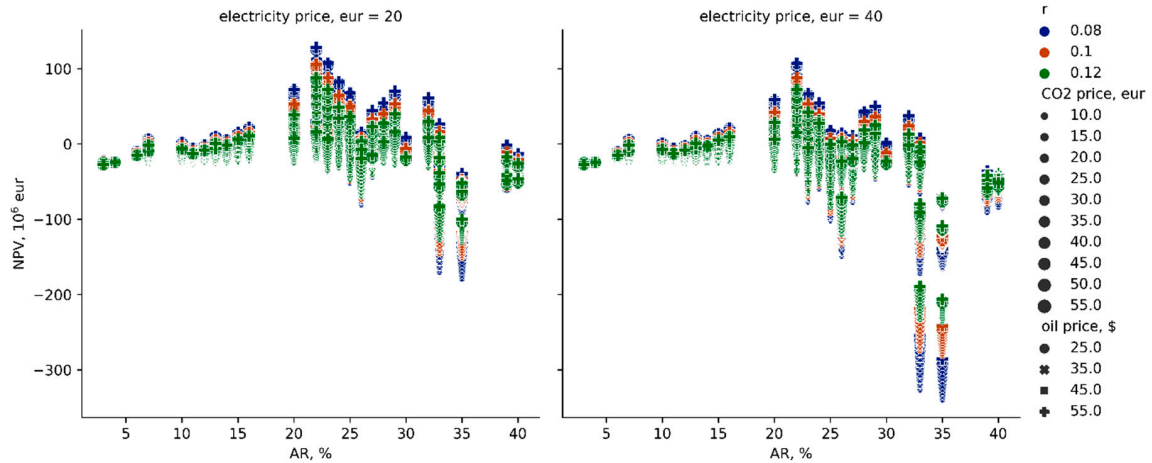


Figure 19. NPV vs. additional recovery after 15 years of CO₂-EOR.

The parameter which relates to additional recovery and retention was the utilization factor (Figure 20). If the only profit from additional oil recovery was observed, the lowest values of UF would be favorable because, in such cases, the highest recoveries could be achieved with the smallest amounts of CO₂ injected. NPV vs. UF showed a qualitatively similar trend to that in NPV vs. retention.

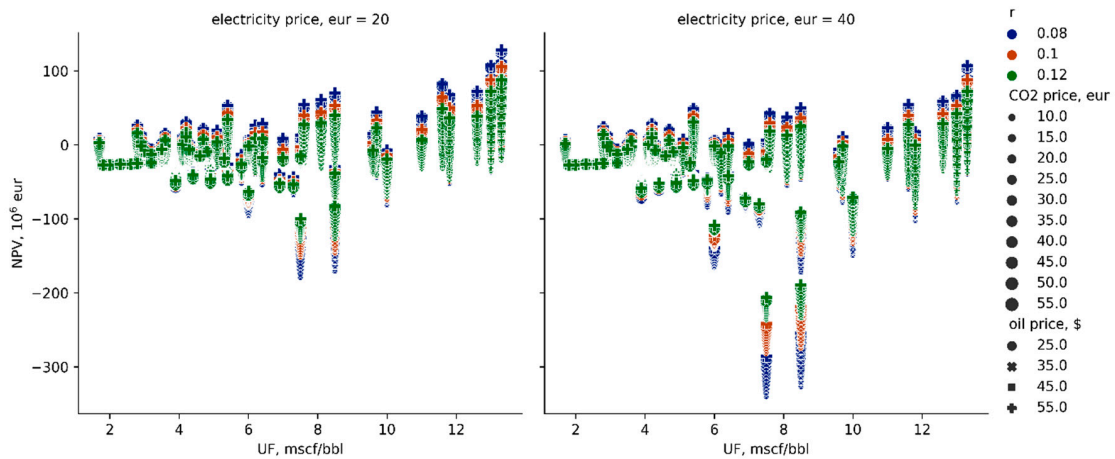


Figure 20. NPV vs. additional recovery after 15 years of CO₂-EOR.

By including different CAPEXs in the analysis, a linear correlation between CAPEX and marginal CO₂ price (Figure 21) was determined.

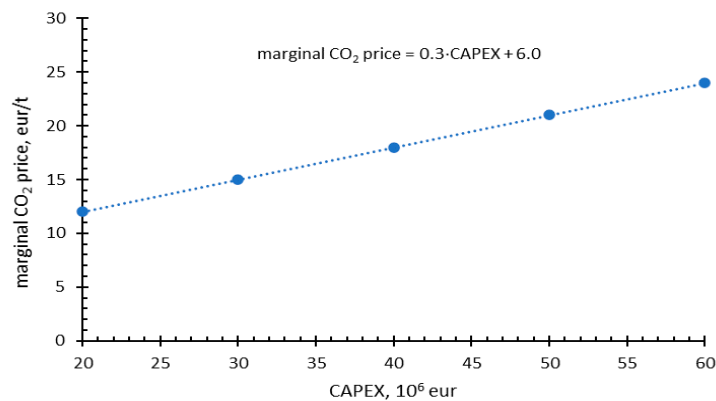


Figure 21. Marginal CO₂ price versus CAPEX.

4. Discussion

Smaller well distance generally resulted in higher retention (Figure 4) because of the higher pressure between the injection and the production wells, which resulted in better mixing conditions and thus preventing viscous fingering of CO₂.

Indicative was that the most injected PV's correspond to the cases with smaller well distance and a 1:2 WAG ratio (Figure 5). Cases with the lowest permeabilities (5 mD) and depths were not among the best cases in terms of injectivity or additional recovery.

If the ratio of the new CO₂ (which should be brought into the system) and recycled CO₂ was observed concerning the additional recovery, it was evident that the highest permeability (250 mD) brought the highest recoveries and greatest depths also brought greater additional recovery while WAG ratio effects were not visible. If the parameters were observed versus time (Figures A5–A9), the distribution of all independent variables was qualitatively the same, but less CO₂ should be brought to the system (more CO₂ recycled) with time.

Storability was inversely proportional to permeability, depth (for the same permeability), and new CO₂ brought to the system (Figures 8 and 9).

The highest operating expenses which were not connected with oil production came from injection costs because of electricity consumption needed for compression (and less significant pumping) of required high quantities of CO₂.

Depending on the applied, because of hourly electricity price fluctuation, electricity price maximum NPV at the end of EOR varies from 0.7 to 16.8 million €. The most cost-effective were those cases with the highest UF, a permeability of 50 mD, a WAG ratio of 1:1 or 1:2, at a depth of 1545 m for both electricity price cases. Generally, more feasible were cases with smaller well distance.

It is not possible to achieve a positive NPV if the CO₂ price is below 15 €/t. Because of the hourly electricity price fluctuation and the special contracts between industrial consumers and the suppliers, the uncertainty of NPV value of the CO₂-EOR will be especially high for the projects with a mediocre performance that might highly depend on electricity price (negative NPV values were more expressed than positive, Figure 18). By observing Figures 18–20, from the similarity of the NPV vs. UF curve (which incorporated both the injected CO₂ and the produced oil) and the NPV vs. retention curve, it was obvious that the oil production would affect NVP less than the CO₂ retention.

Discussed observations were put to statistical analysis of the strength of the relationship between the input and the resulting parameters. The mutual influence of the considered parameters was easiest to show using a diagonal correlation matrix (Figure 22). Values higher than zero indicated a positive correlation between observed parameters, and values less than zero represented a negative correlation. This statistical summary, as the quantitative benchmark of analyzed parameters, was in accordance with previously qualitatively detected relationships observed in Figures 4–20.

A strong negative correlation of storability and additional recovery should be considered together with the positive correlation between depth and additional recovery since greater depths imply higher injection pressures connected with near miscible and miscible conditions for depths over 1545 m, i.e., more CO₂ is dissolved in oil, therefore, more oil is produced, and less CO₂ is retained. That was also the reason for the negative correlation found between storability and additional recovery. The reason for the strong negative correlation between NPV and CO₂ recycled was the fact that if the CO₂ is recycled, then there is a cost of injection, and there is no added value for retained CO₂ (EUA price).

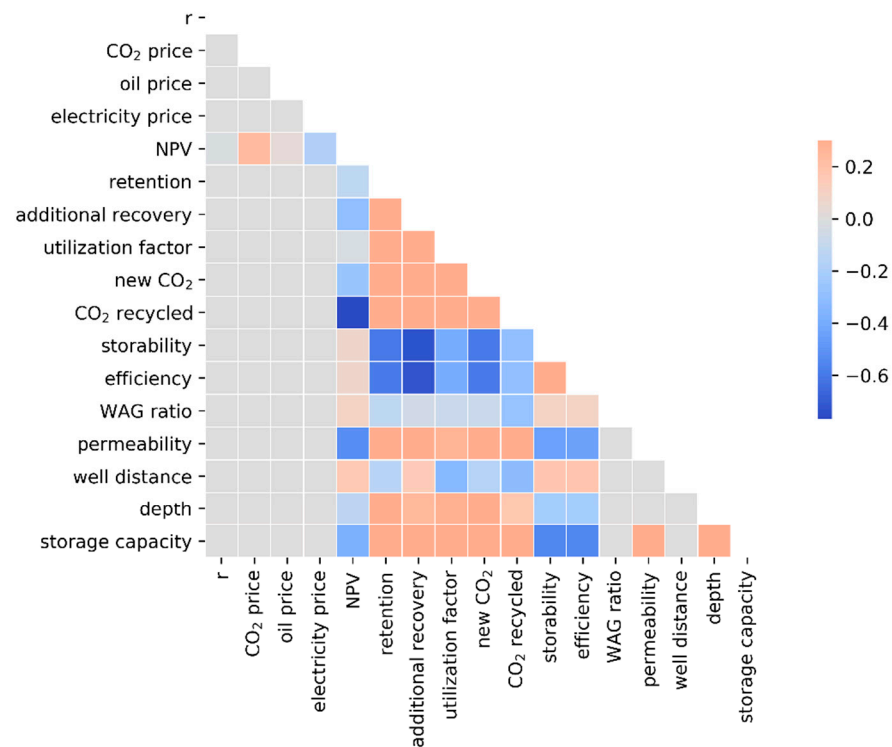


Figure 22. Diagonal correlation matrix.

In general, in terms of NPV, many more parameters gave a negative correlation, and the most intense positive correlation referred to the CO₂ price. As mentioned earlier in the results section, it was not possible to achieve a positive NPV if the CO₂/EUA price was less than 15 €/t and with current prices of CO₂ and oil, it was possible to achieve positive NPV with both maximum and minimum electricity price connected with costs of injection.

The main contribution of this paper is the incorporation of the WAG parameters and the reservoir conditions with the simultaneous economic feasibility assessment, which can serve as some guidelines for multiparameter analysis and optimization of the WAG process. However, considering the number of different values of each observed parameter, this analysis can be improved by introducing more permeability, well distance, and WAG ratio values.

5. Conclusions

The numerical simulation analysis, which consists of 54 different CO₂-EOR injection cases (all combinations of three permeabilities, three depths, two well distances, and three WAG ratios) was taken as a basis for economic assessment (which consisted of three assumptions of interest rate, four CO₂ prices, three oil prices, and two electricity prices), with the final number of cases for economic feasibility analysis $54 \times 36 \times 2 = 3888$.

Considering so many cases was challenging, and it was necessary to create a structured query language (SQL) database of results and inputs and a system using Python code for data post-processing, economic and statistical analysis, and charting.

The following conclusions can be drawn:

1. Retention of CO₂ was affected by the distance between the injection and the production wells; smaller distances between them mean a higher volume of retained CO₂ related to higher permeabilities and depth. The WAG ratio also had an impact on retention: ratios of 1:1 and 1:2 showed bigger retention caused by slug size of gas injection, which means more volume of the CO₂ injected.

2. When considering economic factors, well distance has an important effect on values of NPV. Higher values of NPV were attached to the smaller distances and cases with a permeability of 50 mD. For greater distances, the value of NPV was lower. UF will not be affected by oil and carbon prices, discount rates, and royalties. So, despite the market situation, having a lower UF with positive NPV is an indicator for perspective EOR strategies because of lower expenses for CO₂ recycling.
3. The most optimal case should fulfill the highest NPV and retention and the lowest UF. From the results of this investigation, an optimal case had a permeability of 50 mD, depth ranges between 1545 and 1845 m (from near miscible to miscible conditions), a WAG ratio of 1:2 was the best followed by 1:1. Regarding well distances, the choice should be based on the benefit from a higher NPV (risked to changing oil and carbon prices) with a higher UF or a moderately smaller NPV with a slightly lower UF. In the absence of more simulation results, it can be concluded that there are some optimal (not maximum or minimum) depth and permeability which will give the highest retention, additional recovery, and thus NPV.
4. The maximum correlation value in the diagonal correlation matrix was around 0.2, and the minimum was around -0.6 , i.e., negative correlations were much higher than positive ones, which implies that the CO₂-EOR failure uncertainty will be smaller than the quality of the profit assessments.

The injection of the CO₂ into a reservoir for EOR is economically feasible, and it has been applied for decades. Specifically, for the simulated conceptual models with some previously analyzed properties of oil fields in the Sava depression (Croatia), for the actual prices of carbon and oil, it is economically advantageous to develop new CO₂-EOR projects, and it is clear that including CO₂-EOR in EU ETS for the utilization and storage of CO₂ (CCUS) would make a big difference in terms of CO₂ storage, providing more assets for the application of advanced methods of monitoring and tracking CO₂ over the entire process.

In addition to the CO₂-EOR simulation cases, some pure storage cases were also simulated. For cases with 5 mD permeability, a too high value of pressure is quickly reached in the near-wellbore zone when the CO₂ is injected for storage (there is no oil production and pressure release at production wells), and storage capacities in such cases are not significant and do not account as physically (or economically) feasible. The minimum difference between retention from CO₂-EOR cases and storage capacity from pure storage cases is 1.68 Mt, and the maximum difference is 4.64 Mt (6.76 times more).

Finally, it is important to point out that as more oil is produced during CO₂-EOR (CO₂ utilization project), it is always possible to store more CO₂ after the CO₂-EOR ends, and most of the observed cases (more than 88%) showed the increase of storage capacity even after 15 years of CO₂-EOR.

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Data Availability Statement: Data is contained within the article or supplementary material.

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Conflicts of Interest: The authors declare no conflict of interest.

Appendix A

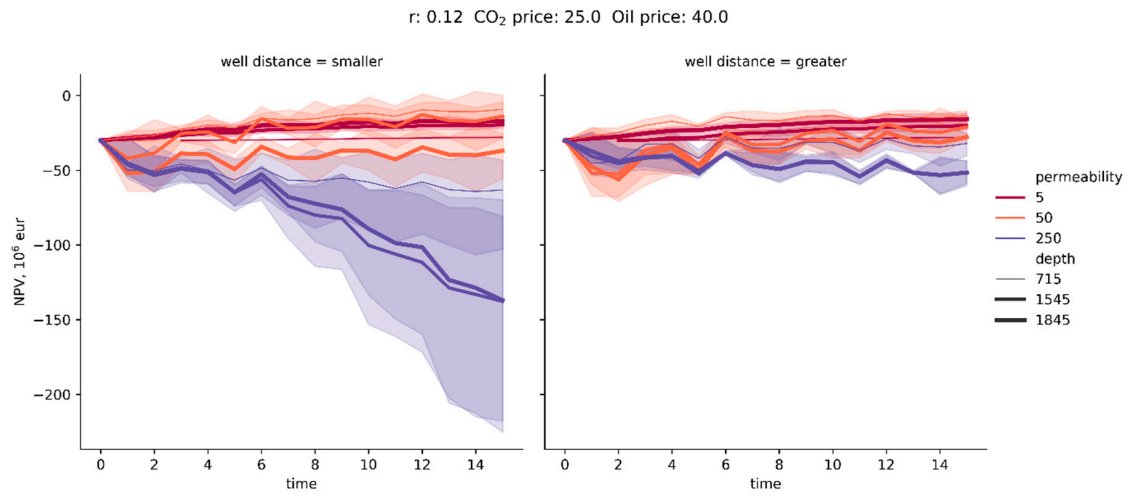


Figure A1. Time versus NPV for the realistic scenario with 40 €/MWh electricity price.

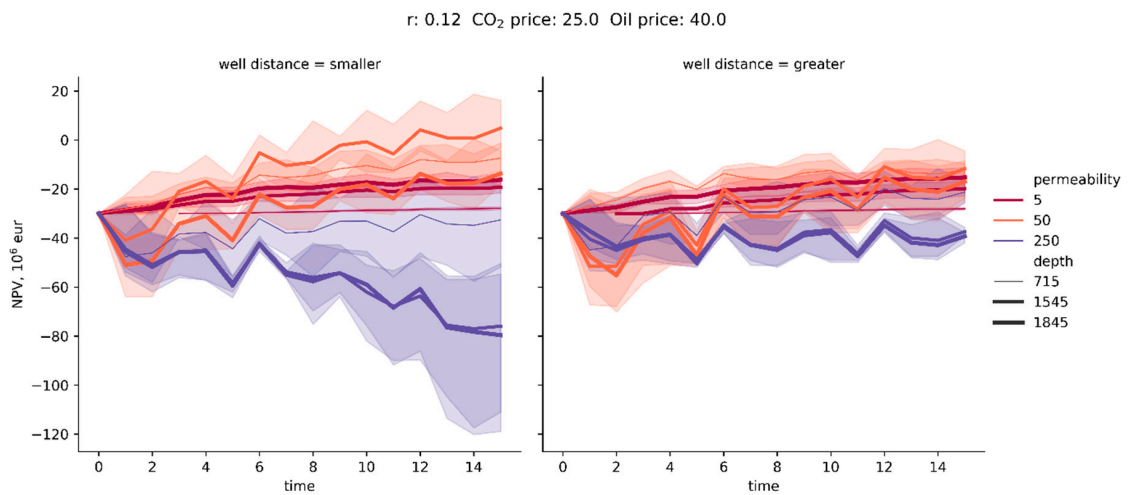


Figure A2. Time versus NPV for the realistic scenario with 20 €/MWh electricity price.

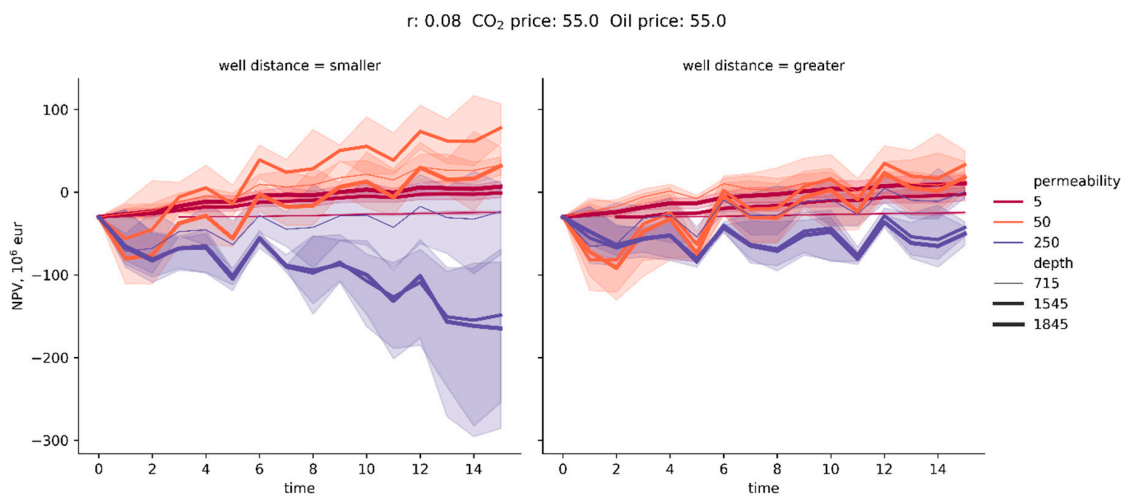


Figure A3. Time versus NPV for the optimistic scenario with 40 €/MWh electricity price.

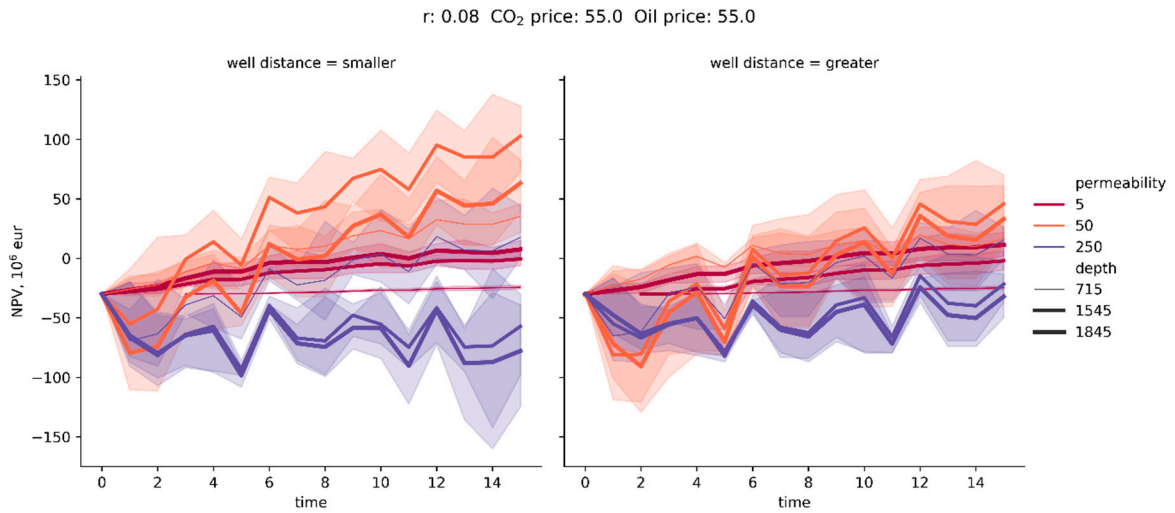


Figure A4. Time versus NPV for the optimistic scenario with 20 €/MWh electricity price.

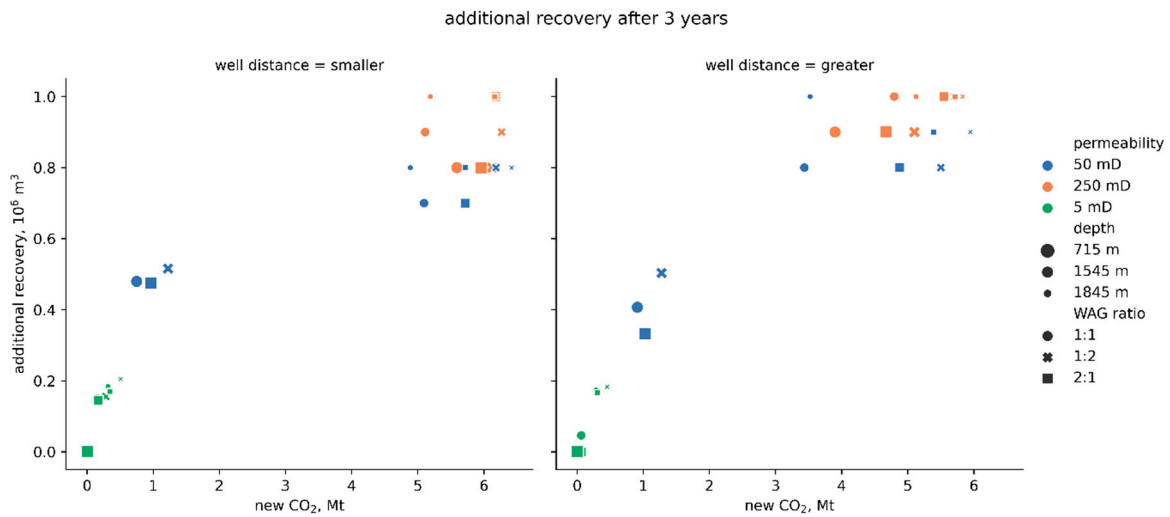


Figure A5. Additional recovery versus new CO₂ after 3 years of EOR.

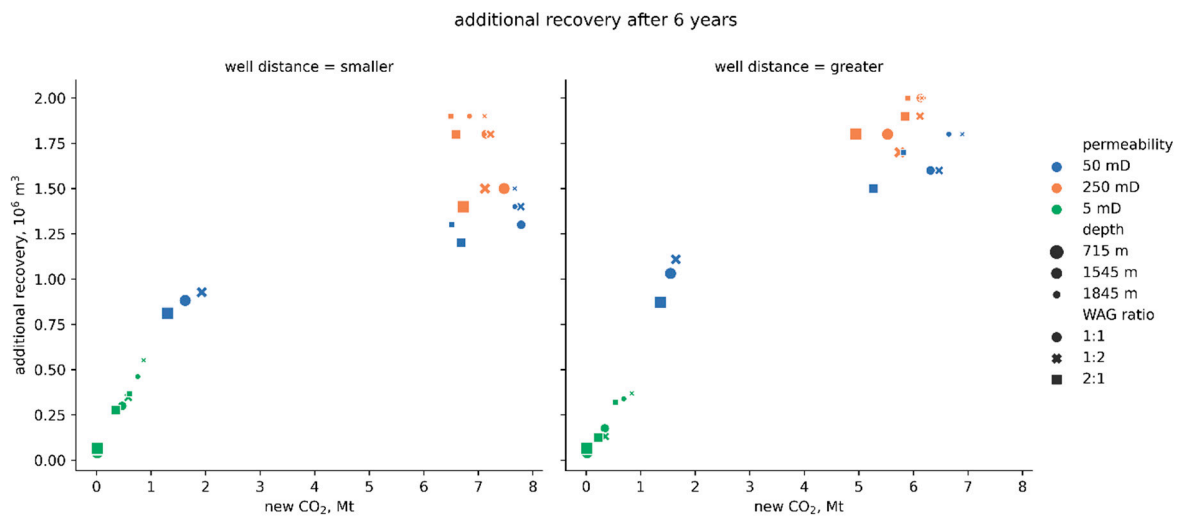
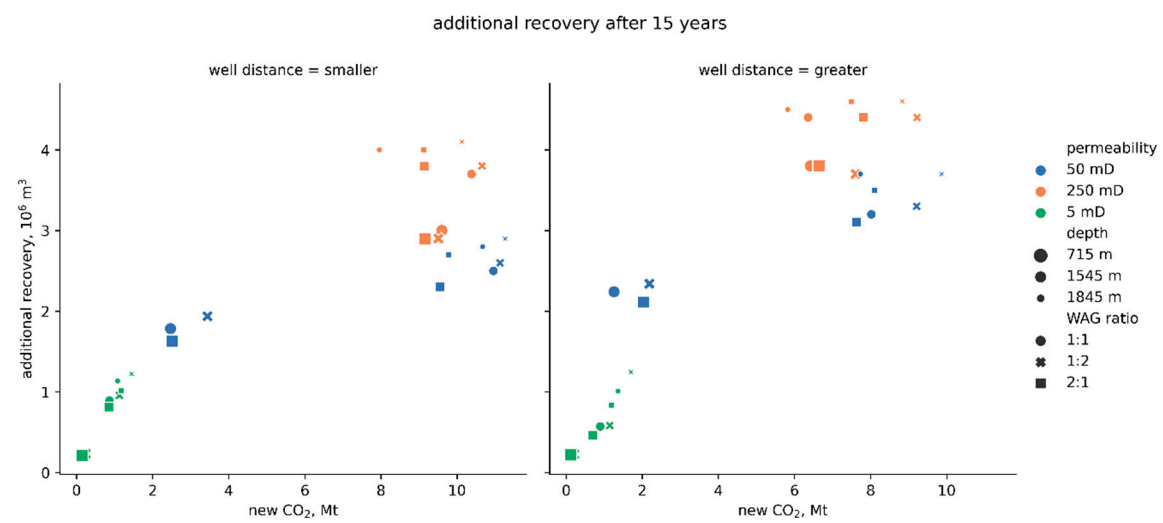
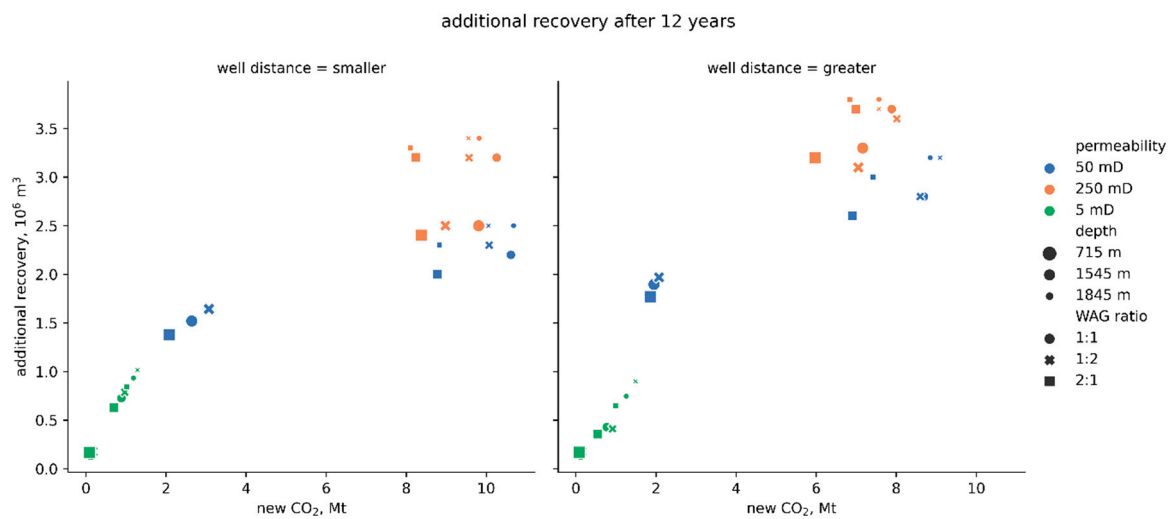
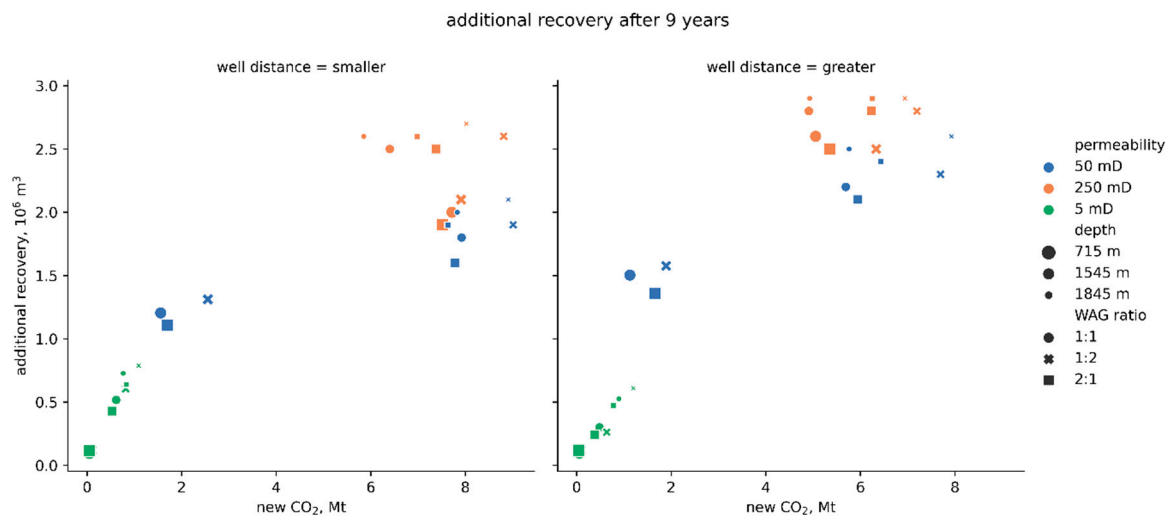


Figure A6. Additional recovery versus new CO₂ after 6 years of EOR.



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