

NEAR EAST UNIVERSITY INSTITUTE OF GRADUATE STUDIES DEPARTMENT OF PETROLEUM AND NATURAL GAS ENGINEERING

MODELING MISCIBLE CO₂/N₂ INJECTION FOR ENHANCED OIL RECOVERY: AIN DAR AREA

M.Sc. THESIS

Noely Sylvanie FOUEDJIO SONFACK

Nicosia

April, 2025

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Approval

We certify that we have read the thesis submitted by Noely Sylvanie FOUEDJIO SONFACK titled "Modeling miscible CO₂/N₂ injection for enhanced oil recovery: Ain dar area" and that in our combined opinion it is fully adequate, in scope and in quality, as a thesis for the degree of Master of Applied Sciences.

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Declaration of Ethical Principles

I hereby declare that all information, documents, analysis and results in this thesis have been collected and presented according to the academic rules and ethical guidelines of Institute of Graduate Studies, Near East University. I also declare that as required by these rules and conduct, I have fully cited and referenced information and data that are not original to this study.

Noely Sylvanie FOUEDJIO SONFACK

10/04/2025

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Noely Sylvanie FOUEDJIO SONFACK

Abstract

Modeling Miscible CO₂/N₂ Injection for Enhanced Oil Recovery: Ain Dar Area

FOUEDJIO SONFACK, Noely Sylvanie M.Sc, Department of Petroleum and Natural Gas Engineering April 2025, 99 pages

This study focuses on the injection of miscible carbon dioxide (CO_2) as a means to improve oil recovery performance. By analyzing various EOR methodologies, notably miscible flooding with CO_2 , which enhances oil dissolution and reduces viscosity, the research aims to assess the impact of nitrogen (N_2) on recovery efficiency. Our study was conducted in the Ain Dar region of Saudi Arabia, utilizing a computer simulator CMG to model the dynamics of oil recovery and evaluate the effectiveness of pure CO_2 flooding compared to mixtures of CO_2 and CO_2 and CO_3 .

The simulation results reveal significant findings regarding oil recovery factors, displacement efficiencies, and cumulative oil production. The natural depletion resulted in a recovery factor of 27.3%, pure CO_2 flooding achieved a remarkable recovery efficiency of 73.8%, while various CO_2 and N_2 mixtures yielded recoveries ranging from 60% to 69%. Notably, pure CO_2 flooding also demonstrated a lower minimum miscibility pressure compared to mixtures of CO_2 with N_2 , contributing to a peak cumulative output of 38 million barrels. These findings underscore the potential of injecting pure CO_2 to optimize oil recovery strategies, while also addressing concerns related to carbon emissions and operational costs.

Key Words: enhanced oil recovery, miscible CO₂ flooding, minimum miscibility pressure, CO₂ sequestration.

Özet

Ghawar Petrol Sahasında Gelişmiş Petrol Geri Kazanımı için Karışabilir CO_2/N_2 Enjeksiyonunun Modellenmesi: Ain Dar Bölgesi

FOUEDJIO SONFACK, Noely Sylvanie Yüksek Lisans, Petrol ve Doğal Gaz Mühendisliği Bölümü Nisan 2025, 99 sayfa

Bu çalışma, petrol geri kazanım performansını iyileştirmenin bir yolu olarak karışabilir karbondioksit (CO₂) enjeksiyonuna odaklanmaktadır. Araştırma, çeşitli EOR metodolojilerini, özellikle de yağ çözünmesini artıran ve viskoziteyi azaltan CO₂ ile karışabilir taşmayı analiz ederek, nitrojenin (N₂) geri kazanım verimliliği üzerindeki etkisini değerlendirmeyi amaçlamaktadır. Çalışmamız, Suudi Arabistan'ın Ain Dar bölgesinde, petrol geri kazanımının dinamiklerini modellemek ve CO₂ ve N₂ karışımlarına kıyasla saf CO₂ taşkınının etkinliğini değerlendirmek için bir bilgisayar simülatörü CMG kullanılarak gerçekleştirildi.

Simülasyon sonuçları, petrol geri kazanım faktörleri, yer değiştirme verimlilikleri ve kümülatif yağ üretimi ile ilgili önemli bulgular ortaya koymaktadır. Doğal tükenme %27,3'lük bir geri kazanım faktörü ile sonuçlandı, saf CO₂ taşkın, %73,8'lik kayda değer bir geri kazanım verimliliği sağlarken, çeşitli CO₂ ve N₂ karışımları %60 ila %69 arasında değişen geri kazanımlar sağladı. Özellikle, saf CO₂ taşkını, CO₂ ile N₂ karışımlarına kıyasla daha düşük bir minimum karışabilirlik basıncı gösterdi ve bu da 38 milyon varillik bir pik kümülatif çıktıya katkıda bulundu. Bu bulgular, karbon emisyonları ve operasyonel maliyetlerle ilgili endişeleri ele alırken, petrol geri kazanım stratejilerini optimize etmek için saf CO₂ enjekte etme potansiyelinin altını çiziyor.

Anahtar kelimeler: gelişmiş yağ geri kazanımı, karışabilir CO₂ taşma, minimum karışabilirlik basıncı, CO₂ sekestrasyonu.

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List of Abbreviations

ASP: Alkaline/Surfactant/Polymer

BHP: Bottom-Hole Pressure

CCS: Carbon Capture Sequestration

CO₂: Carbon-Dioxide

CMG: Computer Modeling Group

EOR: Enhanced Oil Recovery

EOS: Equation Of State

FVF: Formation Volume Factor

GOR: Gas Oil Ratio

GEM: Group Environmental Model

HG: Hydrocarbon Gases

IFT: Interfacial Tension

IEA: International Energy Agency

LPG: Liquefied Petroleum Gas

MMP: Minimum Miscibility Pressure

MCM: Multi-Contact Miscibility

N₂: Nitrogen

OOIP: Oil Originally in Place

OWC: Oil-Water Contact

PR: Peng-Robinson

PV: Pore Volume

PSI: Pounds per Square Inch

PVT: Pressure Volume Temperature

RF: Recovery Factor

ROV: Relative Oil Volume

SRK: Soave-Redlich-Kwong

VLE: Vapor-Liquid Equilibrium

WOC: Water-Oil Contact

CHAPTER I

Introduction

This chapter will provide a foundational background on enhanced oil recovery, with a specific focus on modeling miscible CO₂ enhanced flooding to improve oil recovery. The discussion will include relevant connections between theories related to the research objectives.

Study Background

Over the past 150 years, oil has become the world's primary energy source. Production of this commodity has been rapidly increasing, reaching over 30 billion barrels annually at the beginning of the 21st century. Oil now plays a significant role in global politics and the economy. Although more sustainable energy sources are being developed, hydrocarbons remain irreplaceable for the time being. This reality leads to two main outcomes: either new oilfield discoveries are made, or existing ones are utilized more efficiently. To extract more oil from reservoirs that cannot be tapped using primary and secondary recovery methods, the oil industry employs a set of sophisticated techniques known as Enhanced Oil Recovery (EOR) (Mokheimer et al., 2019). EOR is essential for maximizing the yield from older, depleted oil fields. According to Wang et al. (2017), oil companies can extend the productive life of their oil fields by using EOR methods. These methods significantly increase the total recoverable oil from existing wells. This approach is crucial for meeting global energy demands and optimizing the use of our current resources, ultimately enhancing the oil production process in the long run.

Enhanced Oil Recovery (EOR) aims to optimize the extraction of remaining oil from reservoirs. This process involves creating and testing novel agents, such as water, chemicals, gas, and various liquids with different viscosities, which have shown promise in facilitating smoother oil flow during field applications. EOR consists of three main phases: primary, secondary, and tertiary. During the primary extraction phase, mechanical pumps or the reservoir's natural pressure are used to extract oil. However, this stage typically recovers only a small fraction of the oil available in the reservoir (Malozyomov et al., 2023). According to Alagolorni et al. (2015), secondary recovery techniques can enhance extraction by maintaining reservoir pressure and

dispersing oil through the injection of water or gas. Heavy oil poses a particular challenge due to its high viscosity, making both primary and secondary methods less effective. There are various methods for extracting oil from reservoirs, depending on the energy sources used. The two primary approaches for EOR projects on a commercial scale are thermal and non-thermal techniques. Thermal processes include in-situ combustion, electrical heating, and hot water or steam injection. Conversely, non-thermal processes encompass techniques such as gas and chemical injection. EOR methods, like gas injection, modify the characteristics of the oil, increase its mobility, and make it easier to extract, ultimately leading to greater recovery rates. The primary benefit of EOR is the increased production of oil, which can yield up to 30% more, especially in mature oil fields where reservoir pressure has significantly declined (Al-Wahaibi et al., 2017). Additionally, EOR can contribute to reducing carbon emissions through carbon sequestration efforts. To recover heavy oil from aging reservoirs, petroleum engineers often employ gas injection based EOR techniques. This approach helps maintain pressure, reduce the viscosity of reservoir fluids, effectively displace these fluids, and induce the oil swelling effect. The expansion of reservoir oil occurs when an injected solvent dissolves into the reservoir fluid, making gas injection a longestablished EOR process in the petroleum industry (Willhite, 1998).

Anthropogenic activities have led to an increase in the concentration of carbon dioxide and other greenhouse gases in the atmosphere since the industrial revolution began. The primary cause of this increase is the combustion of fossil fuels for energy production (Filonchyk et al., 2024). Carbon dioxide (CO₂) sequestration is crucial for reducing greenhouse gases, as CO₂ accounts for approximately 64% of the overall impact on the greenhouse effect when compared to other greenhouse gases. To mitigate global warming, it is essential to decrease carbon dioxide emissions. One effective strategy for reducing atmospheric CO₂ is the use of carbon dioxide for enhanced oil recovery (EOR) through reservoir injection. The International Energy Agency (IEA) estimates that there are between 300 billion to 600 billion barrels of oil resources globally that are suitable for the implementation of CO₂ enhanced oil recovery technology (Hall et al., 1983). This represents about 14% to 28% of the world's total recoverable oil resources. Successful examples of carbon dioxide injection for enhanced oil recovery can be found in countries such as China, Russia, the United States, and Canada (Zhao et al., 2021).

One of the most promising techniques among chemical injection in commercial applications is the use of carbon dioxide (CO₂) for enhanced oil recovery (EOR). CO₂ is preferred over hydrocarbon gases (HC) in gas flooding procedures due to its lower cost, higher displacement efficiency, and potential environmental benefits through its disposal in petroleum reservoirs. CO₂ flooding involves injecting carbon dioxide into an oil reservoir to increase its volume and decrease its viscosity, making it easier to extract the oil (Massarweh & Abushaikha, 2022). This method leads to improved oil mobility and higher recovery rates. However, there are challenges associated with CO₂ injection, particularly in heavy oil reservoirs. Common issues include poor sweep efficiency and early CO₂ breakthrough, which can limit the effectiveness of the process. Oil can be displaced by CO₂ through two methods: miscible which is a tertiary recovery and immiscible displacement which is considered as secondary recovery. Miscible processes generally result in higher oil recovery rates compared to immiscible processes. This is because miscible displacement effectively reduces the interfacial tension between the injected fluid and the oil, allowing for more efficient displacement and extraction. Additionally, miscible processes delay gas breakthroughs, which enables the injected gas to remain in contact with the oil for a longer time, ultimately enhancing recovery. In contrast, immiscible processes can impede oil extraction by increasing the oil's viscosity and leaving capillary forces that hold oil within the reservoir. Thus, miscible techniques improve oil recovery by overcoming these challenges.

Problem Statement

The primary and traditional secondary recovery methods, such as waterflooding, applied to the Arab-D reservoir have resulted in low recovery rates and left a significant amount of oil in place. This situation necessitates the exploration of enhanced oil recovery (EOR) techniques. Miscible injection is a promising EOR method that can significantly increase the recovery factor while also reducing carbon emissions through carbon sequestration. In our study, we will investigate and model the injection process of this method. However, the efficiency of miscible injection faces several challenges, including the effects of the fluid's properties and composition, particularly due to the presence of contaminants such as nitrogen (N_2) .

Aim of the Study

This thesis aims to enhance oil recovery in the carbonate reservoir known as the Ghawar oil field by studying the miscibility of CO₂ with oil and evaluating the effectiveness of various CO₂ injection strategies. To achieve this, we utilized CMG GEM to quantitatively assess the viability of Arab-D oil reservoirs for miscible CO₂ flooding. The insights gained from this study enhance our understanding of the phase behavior factors involved in soluble displacement processes, including the prerequisites for a multi-contact miscibility (MCM). It will be demonstrated that the minimum miscibility pressure (MMP) is influenced by the composition of the injected fluid, which in turn affects oil recovery. Techniques for calculating this pressure level using CMG Winprop and empirical relationships are employed. Finally, the results of multiple injection scenarios using carbon dioxide and nitrogen are discussed.

Research Questions

How can injection strategies be optimized to enhance oil recovery? What effects does nitrogen (N_2) have on oil recovery in the Arab-D reservoir? What key factors influence the effectiveness of miscible CO_2 injection? What are the environmental impacts of large-scale CO_2 injection, and how can they be mitigated?

Objectives of the Study

By using simulation studies, the investigation's main objective is to enhance knowledge and design of gas injection with an emphasis on CO_2/N_2 miscibility. In light of this, the following are the study's goals:

Performing the miscible CO_2/N_2 injection in the Ghawar oil field using CMG GEM modeling.

To determine the MMP during the performance of miscible CO_2/N_2 injection using different empirical correlations.

To analyze the effect of impurities such as N₂ on injection strategies for maximizing oil recovery.

To assess the potential environmental risks and benefits of CO_2/N_2 injection.

Significance of the Study

To maximize oil recovery from the Arab-D reservoir, it is essential to implement a miscible approach, as both primary and secondary recovery methods have yielded limited results, leaving a considerable amount of oil still in place. This research aims to address the challenges encountered during miscible CO_2/N_2 flooding, particularly focusing on the effects of CO_2 purity on oil recovery. High purity CO_2 enhances the displacement of oil due to its increased mobility. Conversely, the presence of nitrogen in the system tends to impair oil mobility, displacement efficiency, and overall recovery, limiting the effectiveness of carbon dioxide in the reservoir. However, it is important to note that both carbon dioxide and nitrogen are cost-effective options, which can help lower operational costs, reduce the impact of global warming, and decrease the emission of greenhouse gases into the atmosphere.

Scope

The focus of this research is on tertiary enhanced oil recovery, with a particular emphasis on miscible CO₂ flooding in carbonate reservoirs. The study utilizes CMG as a tool, specifically the CMG GEM simulator, to explore various aspects of this technique. One key area of investigation is the impact of combining CO₂ and N₂ on recovery rates, reservoir compliance, and efficiency of oil sweeps. Additionally, the research delves into how the mixture of CO₂ and N₂ interacts with oil, taking into consideration factors such as miscibility, fluid displacement, extraction efficiency, and the effects induced by nitrogen presence, including changes in oil viscosity and the lowest levels of miscibility achievable. The objective is to refine the miscible CO₂ flooding method to enhance oil recovery processes while prioritizing environmental sustainability. This involves identifying critical parameters and optimal reservoir conditions to maximize sweep efficiency in oil production.

Limitations

Due to the absence of field data and a properly functioning research laboratory during this study, the investigation relies primarily on a comprehensive review of relevant literature to obtain accurate and valuable information. While the CMG simulator effectively models enhanced oil recovery techniques, the results obtained from laboratory experiments such as slim tube tests along with logistical challenges, economic feasibility, and regulatory issues encountered while scaling the approach for field application may not always align with real-world scenarios. This discrepancy can

lead to variations between anticipated and actual performance. Additionally, external factors like market conditions, technological advancements, and shifts in regulatory frameworks may influence the effectiveness and practical implementation of the CO₂ flooding technique, although these aspects may not be directly addressed in this study.

Structure of the Thesis

The thesis is organized into six chapters, each focusing on a specific objective outlined in this study. The first chapter presents the research background, introducing key concepts related to enhanced oil recovery, specifically discussing miscible CO₂ enhanced flooding. Additionally, it outlines the research problem and the objectives aimed at addressing this issue. The chapter also highlights the study's aim, significance, the overall structure of the thesis, and delineates its scope and limitations.

In the second chapter, a comprehensive literature review is conducted. It begins by establishing a theoretical framework, discussing enhanced oil recovery and reservoir simulation with a focus on the CMG GEM model. This section aims to provide foundational research relevant to the project. Furthermore, it includes details regarding the study area's location and geological context.

The third chapter outlines the methodology employed in this research, detailing the steps taken to investigate miscibility and to maximize oil recovery. This includes the methods for data collection, the refined relationships developed for assessing the minimum miscibility pressure (MMP) of CO₂ and N₂, the techniques used for data analysis, and the various injection scenarios designed to achieve the research objectives using the CMG GEM simulation tool.

In the fourth chapter, I present significant findings and analyses related to oil recovery across different scenarios. The primary focus of this chapter is to assess the effects of injecting pure CO_2 versus N_2 on recovery rates, displacement efficiency, cumulative oil production, and other relevant factors.

Chapter five discusses the results regarding the performance of CO₂ injection in the presence of nitrogen, along with an examination of the associated environmental impacts.

In conclusion, the thesis captures the study's findings and provides a summary of the research. It also interprets the results in the context of existing literature and highlights the implications of the findings. Importantly, the conclusion offers recommendations for future research avenues.

CHAPTER II

Literature Review

Enhanced Oil Recovery Overview

Oil recovery refers to the extraction of crude oil from subsurface reservoirs. Traditionally, as depicted in figure 1, the stages of oil recovery are categorized into three main phases: primary, secondary, and tertiary recovery.

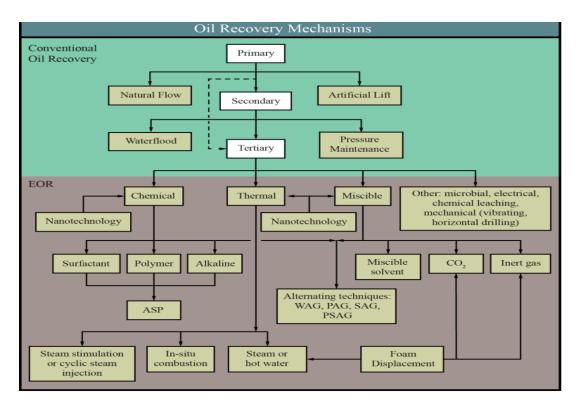


Figure 2.1. Diagram illustrating the steps of extraction of oil and the various strategies utilized in enhanced oil recovery (Energy, 2019).

Primary Recovery: The initial oil production phase relies on the natural energy available within a reservoir to transport oil to the extraction wells. This can involve various natural energy mechanisms such as solution gas drive, gas-cap drive, natural water drive, fluid and rock expansion, and gravity drainage. It is important to note that the classification system does not address the methods used to lift the oil to the surface once it has entered the wellbore.

Secondary Recovery: The second phase of oil recovery typically begins when initial production declines. Common methods of conventional secondary recovery include waterflooding, gas injection, and pressure stabilization. This stage, known as secondary recovery, involves enhancing natural energy by injecting gas or water to facilitate the movement of oil towards the production wells. In this process, gas can be introduced into a gas cap to stabilize pressure and expand the gas cap, or it may be injected into an oil-column well for an immiscible process, based on factors like volumetric sweep efficiency and relative permeability. The immiscible gas injection process involves the introduction of gases, such as nitrogen or carbon dioxide, into an oil reservoir at pressures that are below the minimum miscibility pressure (MMP). In this scenario, the gas does not fully blend with the oil, remaining in a separate phase. This method serves multiple purposes: it helps to uphold reservoir pressure, pushes oil toward extraction wells, and induces a slight swelling of the oil, increasing its volume and facilitating easier recovery. While the gas doesn't completely mix with the oil, it can still help vaporize some of the lighter hydrocarbon components, contributing to additional oil recovery. Overall, this technique is both cost-effective and enhances oil recovery, particularly when achieving miscibility is not an option.

Tertiairy Recovery (EOR): As secondary recovery methods begin to show diminishing returns, tertiary recovery techniques utilize thermal energy, chemical agents, and/or miscible gases to extract additional oil. In some instances, these tertiary processes may serve as a secondary operation. The decision to implement such methods can depend on various factors, including the specific characteristics of the tertiary process, the availability of injections, and economic considerations. For example, it may be more advantageous to bypass the secondary recovery phase if waterflooding could ultimately compromise the effectiveness of subsequent tertiary recovery efforts. These considerations have led to the increased usage of the term "Enhanced Oil Recovery" (EOR) in the field of petroleum engineering, while "tertiary recovery" has become less commonly used. EOR represents a more holistic strategy for managing oil reservoirs, focusing on maximizing oil extraction from existing deposits (Wang et al., 2017).

EOR Classification

Tertiary recovery methods aim to extract the oil that remains in the reservoir after the completion of secondary recovery techniques. Essentially, these procedures focus on addressing the residual oil saturation that persists following secondary recovery efforts. Enhanced Oil Recovery (EOR) encompasses four primary techniques: thermal methods, miscible processes, chemical treatments, and additional approaches such as microbial EOR.

Thermal EOR: Thermal-enhanced oil recovery methods are employed to elevate the temperature of oil, thereby reducing its viscosity through the introduction of heat into the reservoir. These techniques are primarily applicable to heavy, viscous crude oils. The most commonly utilized thermal recovery methods include in-situ combustion and steam injection. Among steam injection techniques, steam flooding, steam-assisted gravity drainage (SAGD), and cyclic steam stimulation (often referred to as huff and puff) stand out as popular approaches.

When steam flooding occurs, the fluids are pushed toward production wells that are positioned strategically by the injection of steam through specific injection wells. SAGD employs a different strategy, involving the heating of the reservoir through a series of parallel horizontal wells. Conversely, cyclic steam stimulation is a method that focuses on a single well, where steam is injected for a designated duration to enhance oil extraction.

Moreover, in-situ combustion entails injecting air or oxygen into the reservoir, where it combines with oil to produce heat, further reducing its viscosity and facilitating its flow to the production wells. Overall, these thermal techniques harness heat energy to enhance oil mobility, thereby improving recovery rates from challenging reservoirs.

Chemical EOR: Introducing specific chemicals into a reservoir to improve oil recovery can alter the properties of the surrounding rock (Malozyomov et al., 2023). The primary aim of chemical flooding, commonly referred to as enhanced oil recovery (EOR), is to boost oil extraction through various methods. These methods include mobility control, which utilizes polymers to reduce the mobility of the injected water,

and the reduction of interfacial tension (IFT) through the application of surfactants and/or alkalis.

In the process of polymer flooding, a water-soluble polymer is introduced to enhance viscosity, thereby facilitating better mobility control during oil displacement. Another technique, known as surfactant flooding, involves injecting surfactants into the reservoir to lower the interfacial tension between oil and water, subsequently enhancing oil recovery. Alkaline flooding also contributes to oil displacement and emulsification by adjusting the reservoir's pH with alkaline substances (Ning et al., 2018). A variation of this method combines the injection of surfactants with alkaline chemicals to optimize the process, while also incorporating polymers for improved mobility control. This combined approach is referred to as an alkaline/surfactant/polymer (ASP) flood.

Miscible Gas EOR: Miscible gas processes refer to methods where the miscibility of in-situ oil and the injected fluid is primarily responsible for the oil displacement's effectiveness. Various gases and liquids can serve as effective miscible agents in both first-contact miscibility (FCM) and multiple contact miscibility (MCM) scenarios. Suitable options include low-molecular-weight hydrocarbons, hydrocarbon mixtures such as liquefied petroleum gases (LPGs), carbon dioxide (CO₂), nitrogen (N₂), and combinations of these substances. Miscible gas injection techniques utilize the distinct properties of gases to facilitate oil movement in order to enhance recovery rates from reservoirs (Wang et al., 2017). A particularly valuable method involves the injection of gases like CO₂, which can be effectively employed in both carbonate and sandstone formations containing oil reserves. This technique is anticipated to gain traction for two main reasons: its potential for sequestering a greenhouse gas and its ability to improve oil recovery through miscibility.

Miscible Gas Process: The main goal of a soluble process is to ensure that the fluid being injected effectively mixes with the oil at the interface where they meet. Since miscibility prevents direct contact between the oil and the displacing fluid, there are no capillary forces present when no interfaces exist. This procedure primarily involves two types: first contact miscibility (FCM) and multiple-contact miscibility (MCM).

FCM: Under specific pressure and temperature conditions in the reservoir, the injected fluid is fully miscible with the reservoir oil in a first-contact miscibility (FCM) process. This method is illustrated in figure 2.

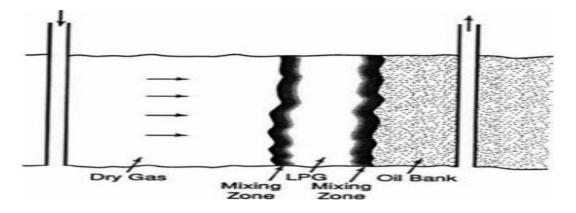


Figure 2.2. FCM process with LPG and dry gas (Willhite, 2018).

The FCM process involves the injection of a comparatively small slug of a hydrocarbon fluid, like liquefied petroleum gas (LPG), which serves to push the oil out. Typically, the size of the primary slug ranges from ten to fifteen percent of the pore volume (PV) (Chequer et al., 2019). Subsequently, a larger volume of a more economical gas with a higher methane concentration, commonly referred to as dry gas, is used to displace the liquefied petroleum gas slug. Additionally, water can be used as a secondary displacing solvent in some situations. The key to the effectiveness of this process lies in the miscibility between the oil phase and the primary slug, which facilitates the mobilization of oil droplets and helps to push them ahead of the primary slug, effectively reducing interactions between the two. Additionally, it is advantageous that the secondary displacing fluid (dry gas) and the primary slug are miscible; otherwise, there is a risk that as the process progresses, the main slug can get caught as a residual phase. Clark et al.(1958) outlined the fundamental requirements for both immiscibility and first-contact miscibility in this context.

MCM: The multiple-contact-miscible (MCM) process is an alternative approach to the conventional miscible process. With this approach, the fluid that was initially injected is immiscible with the reservoir oil. Instead, the process depends on the mass transfer of various components between the two phases within the reservoir, along with their interactions, to modify the composition of the injected fluid. Under

suitable conditions of pressure, temperature, and composition, this alteration can lead to the development of in situ miscibility between the displacing and displaced fluids. The MCM process can be categorized into two primary types: condensing gas displacements (which involve enriched gas) and vaporizing gas displacements (which use lean gas):

- Vaporizing-Gas Displacement Process: The fluid that is typically injected in this process is a lean gas, mainly composed of low-molecular-weight hydrocarbons, particularly methane, although it can include inert gases such as carbon dioxide and nitrogen on occasion. In order to achieve miscibility with the reservoir's existing oil, this procedure involves adjusting the injected gas's composition as it passes through the tank. Essentially, as the gas and oil come into contact, some of the oil's intermediate elements can vaporize into the gas, thereby enhancing the composition of the injected fluid. Under rights conditions, this enhancement can lead to the gas with the modified composition becoming miscible with the oil at specific locations within the reservoir, ultimately resulting in a miscible displacement. However, it is important to note that achieving this miscible condition requires a finite distance, resulting in a transitional zone within the multiple-contact-miscible displacement processes. Laboratory studies indicate that this distance is relatively minimal in reservoir terms (Negahban et al., 1990; Yellig, 1982). Throughout the vaporization process, there may be instances of liquid-phase dropout. The composition of the gas phase can vary between the injection site and the miscible front, as well as behind the front.
- Condensing Gas Displacement Process: In this scenario, the fluid being injected is not merely a dry gas; instead, it contains significant quantities of intermediate components, specifically those in the C₂ to C₆ range. For the process to be effective, these components must condense into the reservoir oil, thus altering its composition. Once this occurs, the injected fluid will mix with the modified oil. A transition zone exists that precedes the miscibility point in this idealized representation of the condensing-type process. This transition is driven by the injected gas moving downstream and continually partially condensing into the oil. The transition zone stabilizes once miscibility is attained. Similar to the behavior observed behind the front in the vaporizing-

gas process, which experiences some liquid condensation, the condensing-gas process shows a minor degree of vaporization occurring in the transition zone downstream of the miscibility point (Zick, 1986).

Miscible CO₂ Flooding

Figure 3 illustrates the process of CO₂ miscible displacement, which entails the injection of a significant quantity of CO₂ to mobilize and recover the remaining oil from a reservoir. Under optimal conditions, this method can lead to a rapid attainment of miscibility within the reservoir. As this transition occurs, oil experiences swelling, and its viscosity decreases. While enhanced miscibility improves oil displacement and boosts sweep efficiency, the injection of CO₂ can also influence interfacial tension by diminishing the capillary forces that trap oil within the pore spaces after water flooding.

In contrast, immiscible CO₂ injection occurs when CO₂ does not fully mix with the oil, remaining as a distinct phase. This technique operates at pressures below the minimum miscibility pressure (MMP), preventing miscibility from being achieved. Although immiscible CO₂ injection lowers interfacial tension, it does not completely eliminate it, which continues to impede oil displacement. Consequently, this approach typically yields lower oil recovery efficiency compared to miscible injection because capillary forces are still at play.

In summary, miscible CO_2 injection is more effective for enhancing oil recovery due to complete mixing and the removal of interfacial tension, whereas immiscible CO_2 injection functions at reduced pressures and results in partial mixing and lower efficiency. Therefore, the amount of recoverable oil increases since the saturation of oil in a miscible state is less than that of the residual oil that remains (Menouar, 2013).

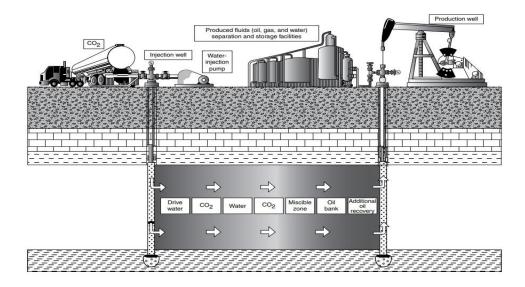


Figure 2.3. CO₂ miscible process (US DOE, 2018).

In the context of miscible CO₂ injection for enhanced oil recovery, miscibility is achieved once the injection pressure exceeds the MMP (Gao et al., 2013). Several factors that can influence both the MMP and the effectiveness of CO₂ injection include reservoir temperature and pressure, the purity of the CO2, the oil composition, fluid properties, and operational conditions, as well as potential CO₂ leakage. To enhance oil recovery rates and optimize sweep efficiency, it is crucial for operators to understand the interactions and flow dynamics of these fluids within the reservoir accurately. Key determinants of CO₂ miscibility involve parameters such as pressure, residence time, and slug size, as well as the fluid characteristics including surface tension, density, viscosity, and purity (Kurdi et al., 2012; Choubineh et al., 2019; Bhatti et al., 2018). Various approaches, including analytical models and empirical correlations, are applied to estimate the MMP. However, acquiring physical measurements can be resource-intensive and expensive, while mathematical modeling may introduce significant errors in the estimated figures. For this study, the Winprop software is employed, utilizing empirical correlation techniques to evaluate the MMP during the CO₂ flooding process.

A simulation model was created by Nasir and Amiruddin (2008) to examine the fluid characteristics' sensitivity to the introduction of miscible CO₂. The effects of various fluid characteristics, including viscosity, formation volume factor, and the density of injected gas and oil, were examined for this aim. According to the data, the characteristics of the recovered oil and injected gas determine how well the miscible

CO₂ injection goes. Furthermore, the final recovery is far more affected by the formation volume factor than by the oil's density. Miscible CO₂ should ideally aid in oil mobility in a number of ways, such as improving microscopic efficiency, decreasing oil density, and lowering interfacial tension (IFT) when in the miscible form.

Al-Abri and Amin (2010) explored how relative permeability relates to interfacial tension and the efficiency of injection displacement in gas condensate reservoirs. To simulate actual reservoir conditions, they conducted experiments using sandstone cores in a high-pressure and high-temperature laboratory environment. The temperature was held at 95°C, and the displacement rate was maintained at 10 cm/h. Their research included displacement experiments at various pressure conditions: immiscible at 1100 and 1200 psi, near-miscible at 3000 psi, and miscible at 4500 and 5900 psi. The core flooding results revealed that pressure was the key factor influencing sweep efficiency. Under near-miscible conditions, only 23% recovery was achieved, whereas miscible flooding resulted in a maximum recovery of 32%, thanks to an improved mobility ratio and delayed gas breakout. The study also highlighted that the characteristics and the phase behavior between CO_2 and condensate played a crucial role in stabilizing the displacement front and enhancing recovery due to the interfacial tension effects during miscible displacement. Similarly, Han et al. (2014) investigated CO₂ injection under miscible conditions using a 2D vertical sandstone model with unstable gravity drainage. Upon achieving miscibility at 100% oil saturation with CO₂ injected from the bottom, they documented an increase in oil production (Bhatti et al., 2019).

In a study conducted by Abbasi et al. (2010), core samples from an Iranian reservoir were subjected to an experimental analysis where they were injected with varying amounts of CO_2 and N_2 gas. The findings revealed that as the injection pressure approached the conditions for miscibility, there was a noticeable increase in the recovery factor. It was observed that the minimal miscibility pressure (MMP) decreased with an increase in the percentage of CO_2 , indicating that higher CO_2 concentrations facilitate the attainment of miscible conditions (Abbasi et al., 2010).

Additionally, Fath and Pouranfard (2013) explored several scenarios of natural reservoir depletion in one of Iran's oil fields. Utilizing a model consisting of 600 grid

blocks, they simulated the behavior of a slim tube at different pressures to determine the MMP. Their results showed that for CO₂ injection, the MMP was approximately 4630 psi (Zene et al., 2019). The simulations indicated that optimal results were 17,000 Mscf/day and 30,000 Mscf/day for the immiscible and miscible injection respectively (Khan et al., 2003). After 20 years of total oil production, the immiscible injection shows a recovery factor of 34.5%, while for miscible injection, it was slightly higher at 36.6%. These results highlighted the greater feasibility of miscible injections, although achieving miscibility in heavy oil reservoirs proved to be challenging at certain points.

A 400 MHz NMR micro-imaging device was used by Yuechao et al. (2011) to investigate the injection of CO_2 into a sand pack. They discovered that the differences in viscosity and density between oil and CO_2 led to the phenomenon of CO_2 fingering during immiscible CO_2 displacement. Consequently, 53% of the oil remained in the sand pack. The study indicated that the sweep efficiency for miscible CO_2 flooding was significantly better. Notably, the CO_2 exhibited low density and viscosity, with a constant flow rate. In contrast to the immiscible injection method, which left 34% of the sand pack containing oil, the findings suggest that miscible injection techniques can significantly enhance oil recovery rates.

CO₂ Enhanced Oil Recovery and Sequestration

As conventional oil reserves become increasingly depleted, the method of carbon dioxide-enhanced oil recovery (CO₂-EOR) has gained significant interest. This technique not only aids in extracting additional oil but also plays a vital role in mitigating global greenhouse gas emissions by sequestering CO₂ deep within geological formations (Kabir et al., 2023). Over the years, the oil industry has effectively utilized this technology, particularly for recovering oil from reservoirs that are approaching their maturity phase. Typically, CO₂ is injected using a continuous flooding method, however, the presence of impurities such as nitrogen (N₂) in the CO₂ can pose challenges, including reduced mobility, viscous fingering, and premature CO₂ breakout, particularly in heterogeneous reservoirs. To improve CO₂ mobility and enhance oil recovery rates, it is beneficial to adjust the injection strategies. Rezk et al.

(2019) highlight that in addition to aiding in oil recovery, CO_2 can also be stored efficiently within reservoirs.

As noted by Rackley (2010), the concentration of carbon dioxide (CO₂) in the atmosphere has significantly increased, rising from 280 parts per million in 1750 to 368 parts per million by the year 2000, and reaching 388 parts per million in 2010. Wang et al. (2011) projected that China's annual CO₂ emissions would escalate from 5.1 billion tonnes in 2005 to 8.6 billion tonnes by 2020. This represents an average annual increase in emissions of approximately 4.5%. The primary driver behind the rising levels of CO₂ is global energy consumption, which is closely linked to climate change and environmental degradation. Hassan et al. (2019) indicate that injecting CO₂ is among the most efficient strategies for enhancing recovery rates. Furthermore, CO₂ that has been combined with oil can be separated and re-injected, offering a storage solution that can significantly reduce greenhouse gas emissions (McLaughlin, 2016). In addition, studies conducted by Wells et al. (2007), Özkilic and Gumrah (2009), and Yang et al. (2010) propose that depleted or mature oil and gas fields with permeable and porous reservoirs may be suitable locations for CO₂ storage.

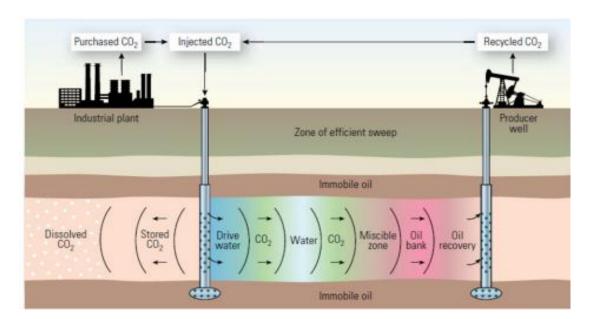


Figure 2.4. CO₂ utilization and sequestration (Aroher and Archer, 2010).

Theoritical Framework

In our thesis, we will establish a theoretical framework that encompasses all pertinent theories and literature related to our research topic, particularly in relation to CMG software workflows.

Reservoir Simulation: The movement and interaction of fluids (oil, water, N_2 , and CO_2) within the reservoir rock are simulated by a reservoir simulation model, which is a mathematical depiction of a subsurface reservoir. To forecast the reservoir's behavior under different circumstances, the model takes into account fluid, geological, and petrophysical characteristics. A simulation model is a simplified representation of a real system that exhibits the key characteristics of the real system or performs aspects that are comparable to it (Jitendra Sangwai & Dandekar, 2022). These computations may be numerical or analytical. By "analytical," we mean that the model's equations could be resolved by mathematical methods similar to those applied to algebraic or differential problems. Analytical solutions are typically defined using "well-known" equations or functions (x^2 , $\sin x$, e^x , etc.) (Heriot-Watt, 1997). A reservoir simulation model serves a variety of functions, such as:

- Production Forecasting: Models aid in the long-term prediction of fluid distributions, reservoir pressure variations, and future rates of oil and gas production.
- Improving Recovery: They support the optimization of recovery strategies, such as CO₂ injection and other enhanced oil recovery (EOR) approaches.
- Risk assessment: Models can detect possible risks and uncertainties related to reservoir management by simulating various scenarios.

We have different types of models including:

• Mathematical Model: By Aziz et al. (1979), a suitable set of mathematical equations must be used to express the physical system to be described. Many assumptions are formed during this procedure. For practical reasons, it is necessary to make assumptions to stabilize the situation. For example, we are forced to use the relative permeability concept even though all reservoir engineers are aware of its shortcomings.

- Numerical Model: The reservoir's mathematical model's equations are nearly
 always intractable using analytical techniques. Approximations must be made
 to describe the equations in a fashion that digital computers can solve. This
 type of equational set produces a numerical model (Sangwai et al.,2022).
- Computer Model: A computer model of the reservoir is a computer program or collection of algorithms created to solve the equations of the numerical model (Sangwai et al., 2022).

GEM Simulator: GEM is a sophisticated compositional simulator designed by the Computer Modeling Group (CMG) that serves the needs of reservoir simulation and petroleum engineering. This simulator features advanced capabilities such as dual porosity modeling, the ability to handle miscible gas interactions, and intricate phase behavior, which collectively enable engineers to analyze fluid dynamics within reservoirs. GEM makes use of established equations of state, including Peng-Robinson and Soave-Redlich-Kwong, to accurately predict the behavior and properties of different phases. Additionally, it applies the quasi-Newton successive substitution method for its nonlinear calculations. The comprehensive input and output data provided by the simulator allow engineers to effectively refine field development strategies and enhance recovery plans.

Related Research

Nitrogen (N₂) presents an intriguing alternative to carbon dioxide in various applications, particularly in enhanced oil recovery (EOR). Its abundance comprising around 78% of the atmosphere makes it a cost-effective option that is also less corrosive than carbon dioxide (Sheng and Soliman, 2013). However, while nitrogen has potential advantages, its utilization has not reached the same level as CO₂ due to specific physicochemical properties. Notably, nitrogen's lower solubility in crude oil complicates its effectiveness in reducing oil viscosity and interfacial tension benefits that carbon dioxide provides more readily.

Prior research illustrates the solubility advantages of carbon dioxide over other gases. For instance, Beecher and Parkhurst (1926) demonstrated that carbon dioxide has significantly greater solubility on a molar basis in 30.2° API oil compared to

natural gas and air. Similarly, Nguyen and Ali (1998) reinforced that carbon dioxide is the most soluble gas in bitumen when compared to methane and nitrogen.

Numerous experimental studies and successful field cases worldwide have showcased nitrogen's potential as an injection gas for enhancing oil recovery. Simulation studies suggest that nitrogen can be effectively employed in regions like Trinidad (Sinanan and Budri, 2012) and southeastern assets (Belhaj et al., 2013). However, it is important to note that the minimum miscibility pressure (MMP) of nitrogen is generally higher than the fracture pressure of the reservoirs, indicating that nitrogen is more suitable for applications at lower pressures where immiscible displacement methods are preferred (Tileuberdi & Gussenov, 2024).

In their exploration of various EOR techniques, Joslin et al. (2017) found that nitrogen flooding is particularly effective for pressure maintenance in reservoirs with permeability lower than 0.03 mD. While there has been significant investigation into nitrogen injection in tight oil reservoirs, research into the feasibility and mechanisms of simultaneous CO_2/N_2 injection remains somewhat limited and merits further study.

The work by Fong et al. (2016) highlighted both opportunities and challenges when utilizing nitrogen during vertical core flooding with 21° API oil. Their findings revealed that using flue gas with approximately 80% nitrogen and 20% carbon dioxide led to earlier gas breakthroughs compared to pure CO₂ scenarios. This resulted in a recovery of only 23% of oil initially in place (OOIP) with the nitrogen-enriched mixture, while pure CO₂ yielded 56% recovery. These insights underline the importance of evaluating optimal gas compositions for enhanced recovery.

Recent research by Lansangan and Smith (2017) on the diffusion characteristics of a 10% nitrogen and 90% carbon dioxide mixture in oil indicated that nitrogen presence decreases the diffusion rate of carbon dioxide by forming a stagnant phase. This phenomenon reduces the interfacial equilibrium concentration of carbon dioxide, hindering mass transfer rates even at lower nitrogen concentrations.

Furthermore, Holm and Josendal (2019) indicated that carbon dioxide alone does not completely displace methane when in contact with oil, while Monger (2019) identified that nitrogen's presence negatively impacts carbon dioxide solubility in oil, unlike sulfur dioxide, which might enhance it.

In a study conducted by Spivak and Chima (2019) on an 82% carbon dioxide and 18% nitrogen mixture, it was found that only a small portion of nitrogen was soluble in Wilmington oil under specific conditions. This reduced the viscosity reduction effectiveness due to diminished carbon dioxide solubility in the presence of nitrogen. Their work also noted earlier gas breakthroughs, lower recovery rates, and a more dispersed compositional front compared to pure carbon dioxide, illustrating the need for continued investigation into the impacts of nitrogen on oil recovery dynamics.

Additionally, Siregar et al. (2020) conducted laboratory experiments assessing nitrogen injection, which revealed that while nitrogen flooding did not lead to substantial increases in oil recovery, other studies indicated that it could yield stable and relatively high production rates. Moreover, Nguyen and Ali (1998) observed that substituting pure carbon dioxide for impure carbon dioxide during the water-alternating-gas (WAG) process adversely affected oil recovery, with higher nitrogen concentrations resulting in losses of up to 10% in recovery rates.

In conclusion, while nitrogen presents distinct advantages as an injection gas for enhanced oil recovery, further research is essential to understand its behavior, optimize its use in conjunction with carbon dioxide, and fully harness its potential in mitigating environmental impacts and improving recovery efficiencies.

Location of the Study Area

The Ghawar oil field located in the Al Ahsa Governorate of Saudi Arabia's Eastern Province, stands as the world's largest conventional oil field. With dimensions of approximately 280 kilometers by 30 kilometers (or about 8,400 square kilometers), Ghawar accounted for nearly one-third of Saudi Arabia's total oil output as of 2018 (Simmons, 2013). Exclusively owned and operated by Saudi Aramco, the state-run oil company, the field stretches approximately 280 kilometers in length, is 30 kilometers wide, and has a thickness of about 90 meters. It is situated on the southern part of the En Nala anticline and is composed of six production regions, arranged from north to south: Fazran, Ain Dar, Shedgum, Uthmaniyah, Haradh, and Hawiyah.

Production at Ghawar began in 1951, following the field's discovery by Saudi Aramco in 1948. The record for the highest daily oil production from a single field

occurred in 1981, reaching 5.7 million barrels per day. While the southern areas, Hawiyah and Haradh, were still undergoing development, overall production began to decline for economic reasons after 1981. As of 2019, the Ghawar field had an estimated production capacity of 3.8 million barrels per day, with remaining proven oil reserves estimated at 58.3 billion barrels of oil equivalent (Gbbl). Our thesis will specifically concentrate on the Arab-D reservoir located in the Ain Dar region (Fink, 2015).

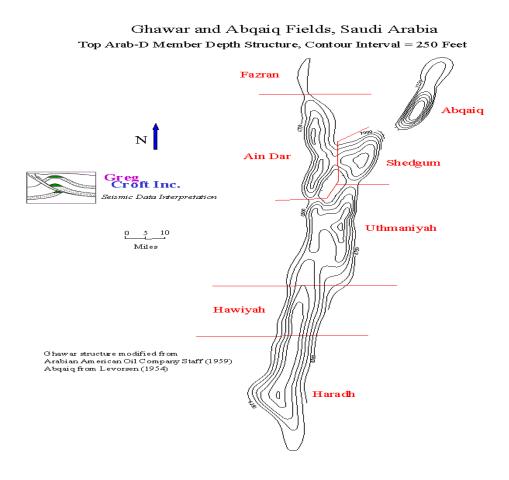


Figure 2.5. Ghawar oil field's map (Saudi Arabia, 2005)

Geological Background: The development of oil reservoirs occurs during the late jurassic to early cretaceous periods, with the Arab-D reservoir serving as the primary source of oil extraction from Ghawar. This reservoir consists predominantly of dolomite, comprising approximately 75% of the upper jurassic dolomites and carbonate formations. It is important to note that dolomites and limestones possess distinct reservoir characteristics, largely due to the processes involved in the diagenesis

of limestone. The Arab-D reservoir is capped by the upper jurassic hith anhydrite, which acts as a seal, effectively preventing vertical migration of oil by trapping vapors of the Arab C-D anhydrite above it (Sultan et al., 2019). The Arab-D reservoir is the source of all the oil produced at Ghawar. Dolomite makes up about 75% of the upper jurassic dolomites and carbonate rocks that make up the Arab-D reservoir. Dolomites and limestones have fundamentally different reservoir properties since these rocks are typically generated during the diagenesis of limestones. Upper jurassic anhydrite, known as hith anhydrite, covers the Arab-D reservoir, sealing it with vapors of Arab C-D anhydrite and preventing oil from migrating vertically (Sultan et al., 2019).

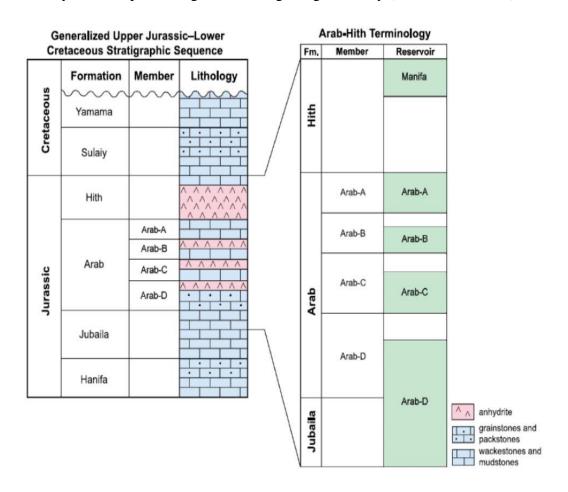


Figure 2.6. General different geological layers, showing in detail the Arab-Hith layers and the location of the simulated Arab-D (Hamimi et al., 2021)

The Arab-D dolomite formation within the Ghawar oil field is responsible for the creation of super-k zones, characterized by exceptionally high permeability. Areas such as Ain Dar qualify as super-k when they exhibit flow rates exceeding 500 barrels per day per vertical foot. The dolomite found in these locales features significant intercrystalline and moldic porosity. When various porous varieties are combined, flow and recovery rates are enhanced. Nevertheless, these super-K areas also present challenges; notably, the wells in these areas demonstrate substantial water flow during secondary recovery processes. Furthermore, the quality of the reservoir diminishes from north to south within Ghawar due to lower porosity and permeability in the southern sections. Additionally, the oil found in the southern region tends to have higher density and sulfur content compared to that in the north, leading to a noticeable decrease in oil quality. The Arab-D dolomite of Ghawar is the cause of super-k or extremely high permeability zones. Additionally, compared to oil in the northern region, the density and sulfur content of the oil in the southern region are higher, resulting in a decline in oil quality (Simmons, 2013).

This thesis investigates the enhanced oil recovery process, focusing on the injection of miscible CO₂ combined with N₂ into the Arab-D reservoir located in the Ghawar oil field through the use of CMG GEM simulation software.

CHAPTER III

Methodology

This project employed a compositional simulation modeling approach utilizing CMG-GEM software, specifically designed to simulate composition reservoirs and understand the intricate interactions among various chemical components. This chapter outlines the simulation process of miscible carbon dioxide (CO_2) for the Arab-D reservoir, focusing initially on the setup and completion of both production and injection wells. The chapter then elaborates on the reservoir fluid PVT data and the Equation-of-State (EOS) modeling methods implemented in Winprop. Additionally, the minimum miscibility pressure (MMP) between CO_2 and the reservoir oil was calculated through the EOS approach. Ultimately, the analysis includes various scenarios of miscible carbon dioxide (CO_2) injection, nitrogen (N_2) injection, and primary production to identify the optimal strategy for enhancing oil recovery and production.

Research Design

This research utilizes a comparative design to evaluate the effectiveness of different injection scenarios of carbon dioxide and nitrogen, aiming to determine the most efficient method. The GEM compositional simulator possesses distinct advantages over IMEX and STARS, particularly in modeling the complex interactions among chemical components. GEM provides flexibility in representing a diverse range of components and phases, and it can accurately model advanced processes, including fluid mixing, chemical reactions, and phase changes. In contrast, IMEX prioritizes simplicity and quick execution for black oil reservoirs, while STARS is proficient in thermal simulations but falls short in accurately modeling chemical interactions. Consequently, GEM is recognized for its capability to deliver a thorough and precise understanding of reservoir dynamics, which is critical for enhanced oil recovery efforts (Agarwal & Liu, 2024).

Using the CMG-GEM compositional reservoir simulator, a three-dimensional (3D) vertical reservoir was constructed. The model comprises 25 grid cells in the x-

direction, 20 in the y-direction, and 6 in the z-direction. The CMG-GEM software facilitated various simulation analyses, including:

- Reservoir Data
- Components
- Rock Fluid Characteristics
- Initial Conditions
- Numerical Data
- Tracer Data
- Well and Recurrent Data

Population of Study/Sample

The study sample was drawn from a reference article titled "Ghawar Oil Field," which was obtained from the Saudi Arabian Oil Company's Global Medium Term Note Programme (Wald, 2018).

Data Collection Tools / Methods and Materials

The primary tool utilized in this study is CMG GEM, which facilitated the development of the simulation model. Numerous iterations were executed through trial and error to identify optimal outcomes. Additionally, Winprop was employed to ascertain the minimum miscibility pressure. To accurately characterize the reservoir fluid sample, the black oil PVT state-based equation was applied. This step is crucial, as incorporating a realistic physical model of the fluid sample into the model is essential before conducting reservoir simulations. Data on compositional characteristics were collected and analyzed in depth to investigate the efficacy of miscible CO2 flooding. An equation of state (EOS) was developed based on information from a traditional pressure-volume-temperature (PVT) study. This study included estimations of fluid properties, two-stage flash experiments, compositional evaluations, differential liberation tests, and separator analyses.

Reservoir Conditions and Fluid Properties

The reservoir model was established using a cartesian grid measuring $25\times20\times6$, resulting in 3,000 grid cells. The hydrocarbon-bearing reservoir spans 14062500 square feet, with I-block and J-block dimensions of 3,750 ft and 3,000 ft, respectively. The model incorporates four production wells (P) and a single injection well (I1), where CO_2 and N_2 are injected under various operational sequences and constraints. This setup is categorized as a standard cuboidal grid model, featuring a five-spot well configuration as depicted in figure 7.

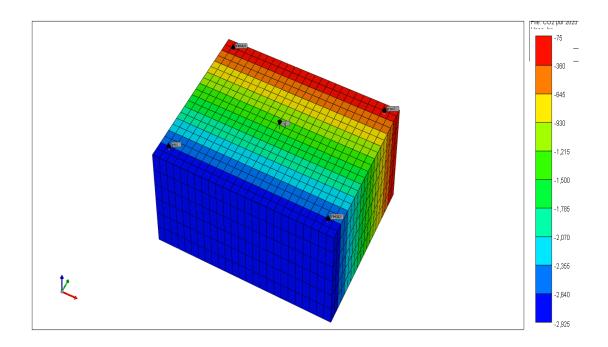


Figure 3.1. Reservoir model (CMG Builder, 2021).

Statistics for the Ain Dar field's Arab-D reservoir fluid indicate an initial gasto-oil ratio (GOR) of 550 SCF/BBL, an average API gravity of 35°API, a reservoir temperature of 215°F and a bubble point pressure of 1,920 psi. The Arab-D formation within the Ain Dar Field exhibits a formation volume factor of approximately 1.34 RB/STB and depths ranging from 5,500 to 6,500 feet. For this study, the formation depth selected is 6,100 feet, resulting in an initial reservoir pressure of 3,212 psi, with an initial water saturation of 0.11.

Table 3.1. General reservoir conditions and fluid properties (Saudi Arabia, 2005)

Properties	Values	
Reservoir temperature (°F)	215	

Bubble point pressure (psi)	1920
Solution gas oil ratio (SCF/BBL)	550
API gravity (°API)	35
Depth (ft)	6100
Oil viscosity (cp)	0.62
Initial reservoir pressure (psi)	3212
Initial water saturation (%)	11
Formation volume factor (RB/STB)	1.34
Permeability (mD)	617
Porosity (%)	19
Injection rate (ft ³ /day)	100000

Compositional Model

Black-oil models, which are often foundational for many reservoir simulations, illustrate how the properties of reservoir fluids fluctuate with variations in temperature and pressure. While these models may suffice for traditional recovery techniques like primary recovery and waterflooding, they are inadequate for enhanced oil recovery (EOR) methods such as miscible gas injection, where fluid composition also plays a significant role in influencing properties. EOS-based methodologies are widely adopted in compositional simulations to depict the phase behavior of reservoir fluids numerically. Each component of an EOS model necessitates critical parameters, including critical temperature, critical pressure, and acentric factors. Including every reservoir component in the compositional model is often impractical due to challenges in quantifying molar fractions and the extensive computational time required for individual simulations. Consequently, components are frequently aggregated into a limited number of pseudo-components to mitigate simulation complexity. To forecast the MMP and the phase behavior of CO₂/N₂/oil interactions, the Winprop software is employed to replicate fluid behavior through the following steps:

- Data collection
- Uploading compositional data into WinProp software
- Exporting the developed EOS model for use in reservoir simulation

For effective EOS-based compositional modeling, accurate EOS modeling is essential. In the petroleum industry, four common cubic equations of state (EOS) are employed:

- Van der Waals EOS (Johannes Diderik, 1873)
- ➤ Soave-Redlich-Kwong (SRK) EOS (Soave, 1972)
- > Peng-Robinson (PR) EOS (Peng & Robinson, 1976)
- ➤ Modified Peng-Robinson (PR78) EOS (Peng & Robinson, 1978)

When employing volume translation, both the Peng-Robinson (PR) and Soave-Redlich-Kwong (SRK) equations of state (EOS) yield comparable predictions for vapor-liquid equilibrium (VLE), as well as accurate estimates for the volumes and densities of vapor and liquid phases. This makes them widely used in the oil and gas sector (Whitson, 2000). In this study, the decision to adopt the PR EOS for fluid modeling was largely based on the acentric factor, which suggests that the acentric factors of the reservoir oil are greater than 0.5. Consequently, we will use the PR EOS to assess gas/oil miscibility and to analyze the phase behavior of the oil reservoir.

PR EOS Development and Modeling: In reservoir engineering, the equation of State (EOS) is a fundamental approach for defining the properties of reservoir fluids and understanding their phase behavior. To accurately simulate the characteristics of miscible CO₂ within a reservoir, it is crucial to mathematically represent these fluid properties. The Peng–Robinson EOS, established by Peng and Robinson in 1976, is widely recognized as the standard in compositional reservoir simulations. This research utilizes the Peng-Robinson EOS to effectively model the fluid properties and phase behavior of the reservoir. The mathematical representation of the Peng-Robinson cubic EOS is detailed in equation 1:

$$P = \frac{RT}{v-b} + \frac{a}{v(v+b)+b(v-b)}$$
 (1)

Where:

T = temperature, ^oR

$$v = specific volume, \frac{ft}{mole}$$

$$R = gas \; constant, \frac{ft^3*psia}{^0R*lb-mole}$$

$$a = attraction \ parameter, \frac{ft^6*psi}{lb^2.mol^2}$$

P = pressure, psi

 $b = repulsion parameter, ft^3/mol$

The parameters associated with attraction, designated as "a" and repulsion referred to as "b," are defined at critical points and depend on both the critical temperature (T_C) and critical pressure (P_C) . This relationship is detailed in equations 2 and 3:

$$b(T_{C}) = \Omega_{b} \frac{RT_{c}}{p_{c}} = 0.07780 \frac{RT_{c}}{p_{c}}$$
(2)

$$a(T_c) = \Omega_a \frac{R^2 T_c^2}{p_c} = 0.45724 \frac{R^2 T_c^2}{p_c}$$
 (3)

With the constants $\Omega_b = 0.07780$ and $\Omega_a = 0.45724$

In regions that are not considered critical points, the parameters a and b can be determined using equations 4, 5, and 6, as presented below:

$$b(T) = b(T_c) \tag{4}$$

$$a(T) = b(T_c) * \alpha(T_r, \boldsymbol{\omega})$$
 (5)

$$\alpha(T_r,\omega) = 1 + (0.37464 + 1.54226\omega - 0.26992\omega^2)(1 - T_r^{0.5})$$
(6)

Where:

 T_r = the reduced temperature, $\frac{T}{T_c}$

Here, the acentric factor ω in equation 6 is specific to each pure component, similar to the critical temperature T_c and critical pressure P_c . For mixtures, various mixing rules are recommended, as outlined in equations 7, 8, and 9:

$$b = \sum_{j} (y_j b_j) \tag{7}$$

$$a(T) = \sum_{i} \sum_{i} [y_i y_i a(T)_{ij}]$$
(8)

$$a(T)_{ij} = (1 - \delta_{ij}) [a(T)_i a(T)_j]$$
(9)

Where:

 δ_{ij} = binary interaction coefficient.

The impact of polar forces on the interactions among system components is incorporated through the binary interaction coefficient in equation 9, which is considered constant regardless of temperature and pressure variations. The parameters a and b from the Peng-Robinson equation of state (EOS) were selected for this study due to their widespread application in characterizing phase behaviors and the physical properties of reservoir fluids. To accurately model the EOS (Equation of State), a range of PVT (Pressure-Volume-Temperature) data is essential. Key parameters include gasoil ratios, oil formation volume factors, relative oil volumes, and the viscosities of oil. These data points play a crucial role in the modeling process.

Compositional Reservoir Simulation: For the sake of simplification, we will disregard factors such as thermal variations, capillary effects, gravitational forces, and diffusion in the subsequent discussion. The foundational mass conservation equation for component i within the two-phase compositional framework is formulated as follows (Chen & Voskov, 2019):

$$\frac{\partial}{\partial t} \left(\Phi \sum_{j=1}^{2} x_{i,j} \rho_{j} S_{j} \right) + \nabla \cdot \sum_{j=1}^{2} x_{i,j} \rho_{j} \mathbf{u}_{j} + \sum_{j=1}^{2} x_{i,j} \rho_{j} q_{j} = 0 , \quad i = 1, ..., N_{c}$$
 (10)

Where:

 S_i = phase saturation, %

 ϕ = porosity of the reservoir, %

 N_c = number of the components

 $x_{i,j}$ = the mole fraction of component i in phase j

t = time, seconds (s)

 ρ_i = molar phase density, mol/m³

 q_i = is the source or sink term of phase j, mol/m³.s

The Darcy velocity $\boldsymbol{u_j}$ is defined as follows (Chen & Voskov, 2019):

$$\mathbf{u_j} = -K \frac{k_{rj}}{\mu_j} \cdot \nabla_p , \qquad j = 1, 2,$$
 (11)

Where:

 $\mathbf{u_i} = \text{darcy velocity, m/s}$

K= absolute permeability, m²

 k_{ri} = relative permeability of phase j

 μ_i = viscosity of phase j, P_a . s

P= pressure, P_a

 ∇_p = pressure gradient, P_a/m

Compositional reservoir simulators rely on the mass or molar balance equations applicable to each component present. Notably, the water equation remains consistent with the previously established black-oil formulation (E. Shtepani, 2007):

$$\nabla \cdot \left(\frac{\lambda_{w}}{B_{w}} \nabla \Phi_{w}\right) + q_{w}$$

$$= \frac{\partial}{\partial x} \left(\frac{\phi S_{w}}{B_{w}}\right)$$
(12)

Where:

 λ_{w} = water mobility, darcy (D)

B_w= formation volume factor of water, RB/STB

 $\nabla \Phi_w$ = Gradient of potential for water, P_a/m

q_w= Source or sink term for water, m³/s

S_w= Water saturation, %

The hydrocarbon equation represents the movement of components i throughout both the oil and gas phases (E. Shtepani, 2007):

$$\nabla . \left(x_i \xi_o \lambda_o \nabla \Phi_o + y_i \xi_g \lambda_g \nabla \Phi_g \right) + q_i = \frac{\partial}{\partial t} [\phi \left(x_i \xi_o S_o + y_i \xi_g S_g \right)]$$

$$(13)$$

Where: $\lambda_p = kk_{rp}/\mu_p$, (p = o, g or w)

 λ_p = phase mobility, darcy (D)

 $\nabla \Phi_p$ = gradient of potential for the phase, P_a/m

 ξ_p = molar phase density, mol/m³

 ∇ = gradient operator, 1/m

 λ_p = Mobility of the phase, $m^2/P_a/s$

 x_i = mole fraction of component i in the oil phase

 S_p = phase saturation

 y_i = mole fraction of component i in the gas phase

 $\frac{\partial}{\partial t}$ = partial derivative with respect to time, 1/s.

Determination of CO₂ Minimum Miscibility Pressure (MMP)

Minimum miscibility pressure (MMP) is defined as the lowest pressure at which multiple contacts between carbon dioxide (CO₂) and oil can achieve miscibility, allowing the two fluids to blend in any proportion without creating an interface, forming a single phase that enhances oil recovery and supports an oil sweep efficiency of around 90% (Holm and Josendal, 1974). In the context of CO₂ flooding, understanding the minimum miscibility pressure is crucial for both the design and operation of the process.

The efficiency of fluid displacement in enhanced oil recovery is closely tied to the MMP. Accurately identifying this pressure enables operators to optimize injection conditions and design surface facilities effectively. A reliable assessment of MMP is vital in commercial applications to determine the feasibility of achieving miscibility between gas and oil under specific reservoir conditions. As noted by Orr & Silva (1987), this critical parameter influences the effectiveness of gas in displacing oil and can be affected by several factors, including temperature, pressure and the fluid composition.

Given the variability of reservoir conditions and the characteristics of the fluids involved, MMP is not a static value; it fluctuates based on the composition of both the gas and the oil. Various empirical correlations exist for calculating MMP, as outlined in equations (14), (15), and (16):

• **Alston** (1985):

$$MMP_{CO_{2(pure)}} = 0.000878(1.8T_R + 32)^{1.06}M_{C_{5+}}^{1.78} (\frac{X_{vol}}{X_{int}})^{0.136}$$
(14)

$$MMP_{CO_{2(impure)}} = MMP_{co_{2(pure)}}.F_{imp}$$
(15)

$$F_{\rm imp} = \left(\frac{87.8}{T_{\rm cm}}\right)^{1.935} \frac{(87.8)^{1.935}}{(7cm)^{1.935}}$$
(16)

Where:

T_R refers to the temperature of the reservoir measured in degrees Fahrenheit (°F).

 $M_{C_{5+}}$ denotes the molecular weight associated with the heavier fractions of fluids, g.mol⁻¹.

 X_{vol} indicates the mole fraction of volatile substances, including nitrogen (N_2) and methane (CH_2) , mol.mol⁻¹.

 $X_{\rm int}$ represents the mole fraction of intermediate compounds, specifically those within the carbon range of C_2 - C_6 .

 T_{cm} is the weight-averaged pseudocritical temperature of the mixture, also expressed in degrees Fahrenheit (°F).

• Glaso (1980): Glaso undertook both correlational and experimental studies focused on oil reservoirs located in the North Sea. According to the findings by Chen et al. (2020), the study utilized established graphical relationships, which played a crucial role in forecasting minimum miscibility pressure (MMP). The study identified

two distinct types of correlations that were determined by the proportion of the intermediate compounds present.

For intermediate mole fractions of C_2 - C_6 exceeding 18 mol, the MMP can be determined using the following formula:

$$\begin{aligned} \text{MMP} &= 5.58657 - 0.02347739 \times \text{M}_{\text{C}_{7+}} + [1.1725 \times 10^{-11} \times (\text{M}_{\text{C}_{7+}})^{3.73} \times \\ & e^{786.8*\text{M}_{\text{C}_{7+}} - 1.058}] \times (1.8\text{T}_{\text{R}} + 32) \end{aligned} \tag{17}$$

Conversely, for mole fractions of C₂ to C₆ less than 18 mol, the MMP is given by:

$$\begin{aligned} MMP &= 20.33 - 0.02347739 \times M_{C_{7+}} + [1.1725 \times 10^{-11} \times (M_{C_{7+}})^{3.73} \times \\ e^{786.8*M_{C_{7+}} - 1.058}] \times (1.8T_R + 32) - 0.836 \times F_R \end{aligned} \tag{18}$$

 F_R indicates the mole percentage of intermediate components (%).

In our research, we employed compositional simulation techniques to ascertain the MMP. The PR EOS has shown to be highly effective for modeling reservoir fluids and is compatible with the CMG software, specifically Winprop, which facilitates the calculation of minimum miscibility pressure (MMP). Winprop utilizes three distinct methods to accurately assess MMP:

- Multiple Mixing Cell (Ahmadi & Johns, 2011): This method simulates a series of mixing cells where various proportions of oil and injected gas are mixed. By examining the phase behavior at different pressures and compositional variations within each cell, we can pinpoint the pressure at which the oil and gas become miscible. One key advantage of this approach is the comprehensive phase behavior data it yields, making it particularly effective for complex mixtures.
- **Tie Line** (Chen & Cheung, 1983): The Tie Line method employs a ternary diagram to visualize the equilibrium compositions of the oil and gas phases. Drawing a tie line on this diagram allows us to find the intersection with the phase envelope, which indicates the pressure at which miscibility occurs. This method is appreciated for its simplicity and speed, making it suitable for systems with clear phase behavior.
- **Cell-to-Cell Simulation** (Dipesh Niraula et al., 2024): This technique simulates the interactions between adjacent cells in a reservoir model. By assessing phase behavior at the interfaces of these cells under various pressure conditions, we

can ascertain miscibility. While this method offers a realistic depiction of reservoir dynamics and fluid flow, it demands considerable computational resources.

Ternary Diagram

To determine phase equilibrium within a pressure-volume cell, we carry out calculations following several contacts between gas and oil. Research by Zick (1986), Johns et al. (1993), and Cook et al. (1969) suggest that multiple contacts may be required to generate phase equilibrium data, from which a pseudo-ternary diagram is created. According to Kanatbayev et al. (2015), both forward and backward contact processes can be analyzed at specific temperatures and compositions across a defined pressure range. Figure 8 illustrates a typical ternary diagram, which, as described, represents three components. The points on this diagram correspond to 100% composition of each constituent, typically including heavier compounds (C_{7+}) and intermediates (C_2 - C_6), with the injection gas represented at the apex. The plait point, marked as point P, is where the dew and bubble point curves intersect, providing insights into composition enrichment conditions, as noted by Mihçakan et al. (1993). The ternary diagram displays several lines and points that convey critical information.

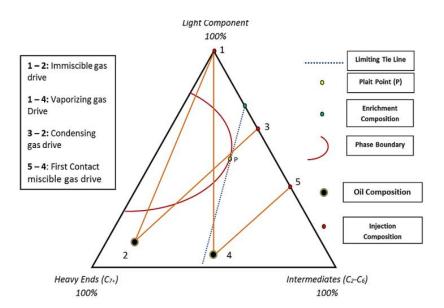


Figure 3.2. A typical representation of a ternary diagram (Adekunle, 2007).

The envelope depicted in the diagram outlines a two-phase zone, where the sample composition is illustrated as a single point. If any line crosses the two-phase envelope, it indicates immiscibility between gas and oil. On the other hand, if the lines stay outside the phase envelope, it implies complete miscibility between gas and oil.

Reservoir Components

To establish the pressure, volume, and temperature properties of the system, we identified ten components within the reservoir. Each component is characterized by its critical molecular weight, critical temperature, and critical pressure. The viscosities of these components, along with the densities of oil and water, were assessed using Winprop software.

Table 3.2. components properties (generated by CMG WINPROP, 2021).

No.	Component	НС	Pc (Atm)	Tc (°K)	Acentric	MW
					fact.	
1	CO ₂	3	72.8	304.2	0.225	44.01
2	CH ₄	1	45.4	190.6	0.008	16.043
3	C_2H_6	1	48.2	305.4	0.098	30.07
4	C_3H_8	1	41.9	369.8	0.152	44.097
5	IC_4	1	36	408.1	0.176	58.124
6	NC_4	1	37.5	425.2	0.193	58.124
7	IC ₅	1	33.4	460.4	0.227	72.151
8	NC ₅	1	33.3	469.6	0.251	72.151
9	FC_6	1	32.46	507.5	0.27504	86
10	C ₇₊	1	15.284	723.823	0.6609	215

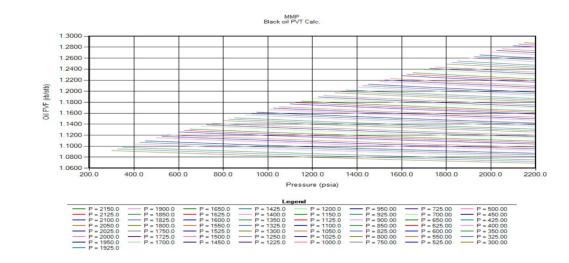


Figure 3.3. The relationship between the oil Formation Volume Factor (FVF) and pressure (CMG Builder, 2021).

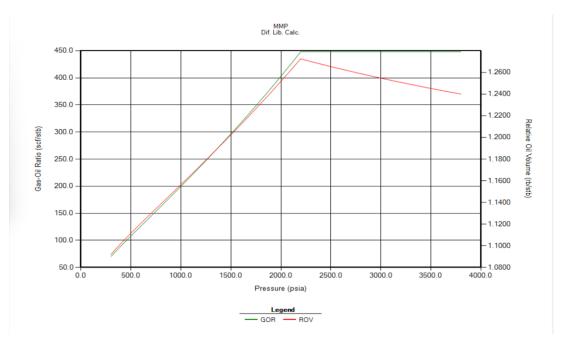


Figure 3.4. The correlation of Gas Oil Ratio (GOR) and Relative Oil Volume (ROV) with Pressure (CMG Builder, 2021).

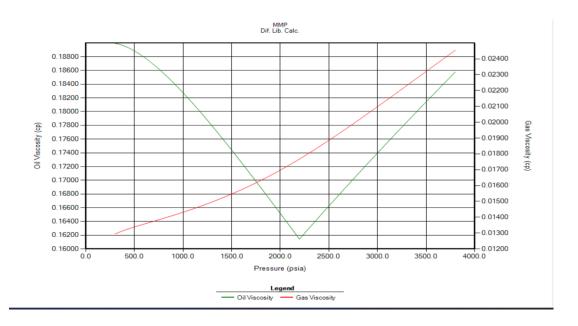


Figure 3.5. The dependence of oil and gas viscosity on pressure (CMG Builder, 2021).

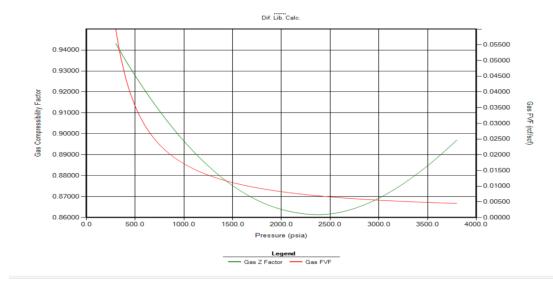


Figure 3.6. The gas FVF and the gas compressibility factor in relation to Pressure (developed using CMG Builder, 2021).

Rock Fluid

In this study, the PVT module of the Winprop software was utilized to determine fluid properties at a temperature of 215°F and a reservoir pressure of 3212 psi. Subsequently, the pressure-temperature phase diagram was generated, illustrated in Figure 14, which depicts the bubble point pressure and the phase behavior of the reservoir oil.

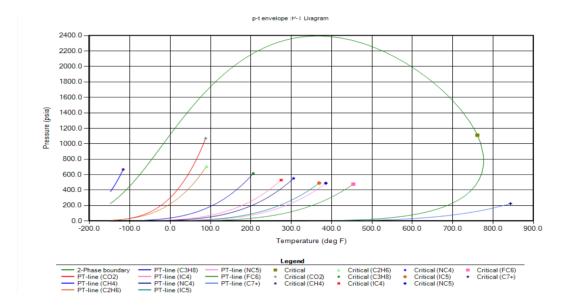


Figure 3.7. Schematic two-phase diagrams of the oil composition used in this study (CMG Builder, 2021).

By utilizing a suitable scaling factor, it is possible to obtain representative values for capillary pressure and relative permeability corresponding to specific pressure and composition conditions. Generally, the miscible displacement for a specific rock section will fall within these established ranges. It is important to note that the phase potential and mobility outlined in equation (13) are not influenced by interfacial tension. There are multiple methods to utilize interfacial tension as a scaling factor for capillary pressure and relative permeability during miscible displacements:

> Relative permeability to gas and oil phase:

$$k_{rg} = \sigma^* k_{rg} |_{\sigma_{max}} + (1 - \sigma^*) k_{rg} |_{\sigma_{min}}$$
 (19)

$$k_{\text{rog}} = \sigma^* k_{\text{rog}} |_{\sigma_{\text{max}}} + (1 - \sigma^*) k_{\text{rog}} |_{\sigma_{\text{min}}}$$
(20)

Where:

 $k_{\mbox{\scriptsize rog}} = \mbox{\scriptsize oil}$ and gas phase relative permeability, mD

S_{or}= residual oil saturation, %

> Capillary pressure:

$$P_{c} = \sigma^{*} P_{c} |_{\sigma_{max}} + (1 - \sigma^{*}) P_{c} |_{\sigma_{min}}$$
 (21)

Where
$$\sigma^* = \frac{\sigma - \sigma_{\min}}{\sigma_{\max} - \sigma_{\min}}$$
 (21.1)

This dimensionless parameter is used to adjust the empirical bounding values. The interfacial tension σ can be determined by employing the modified Macleod-Sugden correlation for multi-component mixtures, a method originally suggested by Weinaug and Katz in 1943:

$$\sigma = \left[\sum_{i} P_{i} \left(x_{i} \frac{\rho_{o}}{M_{o}} - y_{i} \frac{\rho_{g}}{M_{\sigma}}\right)\right]^{\theta}$$
(22)

In this context, σ represents surface tension measured in dynes/cm, while ρ denotes the phase density expressed in g/cm. The molecular weight of the phase is given by M, and the mole fraction of component i in the liquid phase is indicated as x_i , with y_i representing the mole fraction of the same component in the gas phase. The pressure of component i, P_i , can be determined using established correlations. The exponent θ , which is slightly less than 4, was initially introduced by Macleod and Sugden.

In order to accurately model fluid movement in reservoirs, it is essential to comprehend the relative permeability curves that correspond to the equilibrium state of fluids at both high and low interfacial tension during displacement processes. These graphs can serve as the defining bounding relative permeability curves in the simulation model.

At a water saturation of 0.25, the relative permeability of water (k_{rw}) starts to increase in the oil-water phase, as seen in Figure 14, represented by the red line. Conversely, the relative permeability of oil-water (k_{row}) diminishes, ultimately reaching zero at a saturation level of 0.58, as shown by the blue line. Figure 15 depicts the relationship between relative permeability and liquid saturation. This plot indicates that the relative permeability of gas decreases over time, achieving zero at a liquid saturation of 0.86. In contrast, the relative permeability of oil-gas (K_{rog}) shows a declining trend from its initial saturation level until it reaches 0.45.

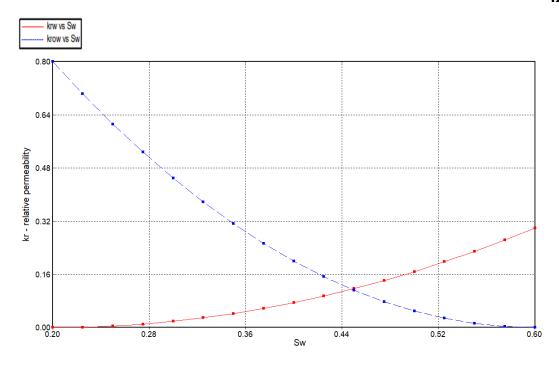


Figure 3.8. Relative permeability versus water saturation (CMG Builder, 2021).

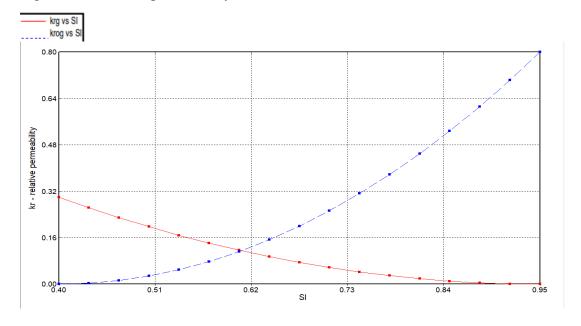


Figure 3.9. Relative permeability versus liquid saturation (CMG Builder, 2021).

Fluid Contacts and Initialization

To accurately validate the simulation, it is essential to establish the initial conditions, which include the depths of fluid contacts (referred to as the oil-water contact depth), as well as the reference depth, pressure, and phase saturation of the grid cells. In this study, we input the oil-water contact (OWC) depth, the water-oil

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contact (WOC) depth, the pressure at the reservoir's reference depth, and the reference depth itself to ascertain the initial state of the reservoir model. The initial pressure was set at 3212 psi, with a reference depth of 6100 feet and an initial WOC depth of 6600 feet. For numerical controls, the standard temperature variation was fixed at 18 °F, while pressure variation was set at 145 psi.

Production and Injection Well Constraints

For the simulation constraints, the producing well was limited to a minimum bottom-hole pressure (BHP) of 28 psi and a maximum surface oil rate of 15000 bbl/day. The injectors were constrained at a maximum surface gas rate of 100,000 SCF/day and a bottom-hole pressure of 2000 psi. Each layer was perforated, enabling various solvents to be injected along the model's flank, while producers were strategically positioned around the injectors to optimize the oil transfer process. The CMG-GEM 2021 simulator completed the initial simulation utilizing the natural depletion of the Arab-D reservoir, along with various scenarios involving different injected fluids, to maximize recovery. The comparison case includes:

- Case 1: Natural depletion
- Case 2: 100% pure CO₂
- Case 3: Impure CO₂, which included
 - \circ 75% CO₂ + 25% N₂
 - \circ 50% CO₂ + 50% N₂
 - \circ 75% N₂ + 25% CO₂
- Case 4: 100% pure N₂

Data Collection Procedures

To thoroughly analyze the findings, it is crucial to consider several factors, including the study duration, oil production rates, cumulative oil output, overall field recovery, and the total amounts of ${\rm CO_2}$ and ${\rm N_2}$ captured. Further details on these aspects will be discussed in the next chapter.

Data Analysis Plan

The research study will follow a structured approach, outlined in the following steps:

- A. Initially, we will compile data from multiple sources related to the Ghawar oil field to establish a thorough dataset that enables detailed investigation.
- B. Utilizing CMG GEM software, a compositional reservoir simulation model will be developed, while CMG Winprop will be employed to determine the MMP by evaluating the miscibility levels between CO₂/N₂ gas and oil through established correlations.
- C. A range of simulation scenarios will be developed to evaluate the impacts of injecting pure miscible fluids as compared to alternative mixtures of N₂. This comparative analysis will assist in identifying the most effective strategy for enhancing oil recovery significantly.
- D. Finally, an assessment will be conducted regarding the potential environmental implications of ${\rm CO}_2$ injection, focusing on issues such as groundwater contamination, risks associated with leakage, and the overall capacity for ${\rm CO}_2$ storage.

CHAPTER IV

Findings and Discussion

This chapter outlines the findings derived from the collected data. After incorporating the reservoir data into the CMG GEM software, we executed the simulation, which yielded significant results. Our research initially aimed to evaluate the efficacy of miscible carbon dioxide (CO_2) injection to enhance oil recovery. The outcomes from the miscible CO_2 injection were scrutinized and compared to those obtained from a mixture with nitrogen (N_2) , allowing us to assess the influence of impurities on key performance metrics. These include the average reservoir pressure, oil recovery factor, cumulative oil production, oil production rate, and carbon capture.

Case 1: Natural Depletion

In this scenario, we examined how the reservoir behaves under natural depletion alongside projected recovery rates, utilizing the GEM software for simulation. The model parameters included a maximum flowing bottom hole pressure of 2000 psi, establishing this case as the base case scenario. Without an injection well and supported solely by four production wells located at the reservoir's periphery, this scenario exemplifies natural production.

The simulation results, spanning a period from 2000 to 2025 (25 years), indicated that the oil recovery factor was 27.3% of the original oil in place (OOIP). By the end of 20 years, cumulative oil production reached 10 million barrels. During the first year, the reservoir exhibited an oil production rate of 8,745.14 barrels per day, alongside a decline in reservoir pressure to 172.5 psi. The graphical representations of the discussed outcomes are displayed in the following figures:

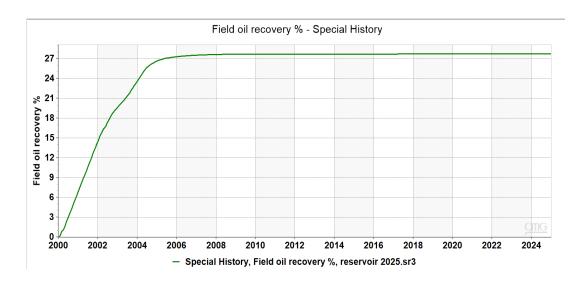


Figure 4.1. Oil recovery factor vs time for base case scenario (generated by CMG Results, 2021).

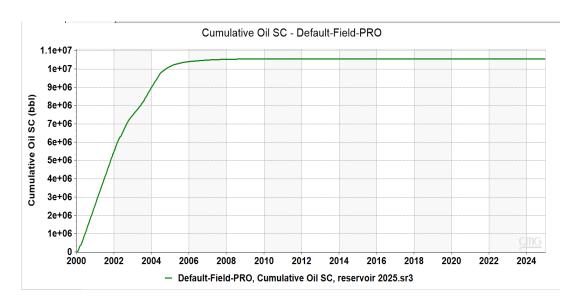


Figure 4.2. Cumulative oil vs time for base case scenario (generated by CMG Results, 2021).

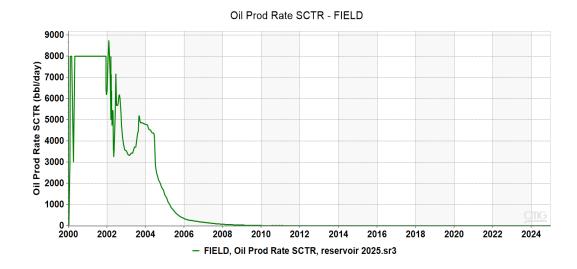


Figure 4.3. Oil production rate vs time for base case scenario (generated by CMG Results, 2021).

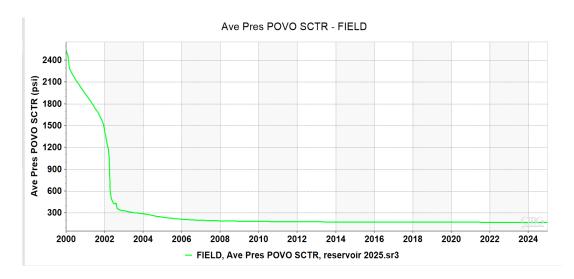


Figure 4.4. Average pressure povo SCTR vs time for base case scenario (generated by CMG Results, 2021).

Case 2: Miscible Pure CO₂ Flooding

In this case, we examined the injection of 100% CO₂ into the reservoir. The method of miscible gas injection involves introducing gas into the reservoir at pressures above the miscibility pressure to enhance oil recovery (El-Hoshoudy & Desouky, 2018; Fath & Pouranfard, 2014). For this scenario, 100% pure CO₂ was injected, and simulations were conducted using a compositional model in CMG GEM 2021. The correlations employed, including multiple mixing, cell-to-cell, and tie lines,

are summarized in the table below, which displays the various pressure values at FCM and MCM for the miscibility of carbon dioxide with oil, given an initial reservoir pressure of 3212 psi.

Table 4.1. MMP for pure CO₂ (generated by CMG Winprop, 2021).

Case	Cell to Cell	Tie Lines	Multiple Mixing
Pure CO ₂	2575 psi	1950 psi	2623 psi

This reservoir is modeled as a standard cuboidal grid with five spot wells. The process operated at a rate of 100,000 ft³/day, utilizing four producing wells (P) and one CO₂ injection well (I1). The miscible gas injection commenced with a maximum bottom hole pressure constraint of 2000 psi. Compared to the base case, the pure CO₂ injection achieved a higher oil recovery factor of 73.8% and a cumulative oil production of 38 million barrels. Additionally, the reservoir pressure decreased to 1000 psi, and the maximum oil production rate at the end of the 25 years was determined to be 60,000 bbl/day. These results were facilitated by the implementation of an injector well with the optimal bottom-hole pressure and gas injection rates, or the surface gas rate (STG), using pure CO₂ as the component fluid. The subsequent figures illustrate the oil recovery factor, cumulative oil production, oil production rate, and average pressure pore volume.

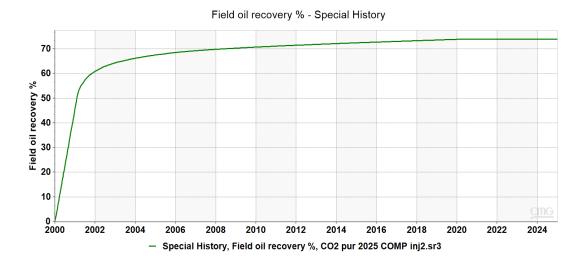


Figure 4.5. Oil recovery factor vs time for 100% CO₂ injection (generated by CMG Results, 2021).

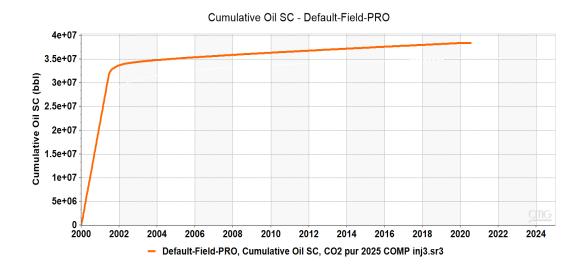


Figure 4.6. Cumulative oil production vs time for 100% CO₂ injection (generated by CMG Results, 2021).

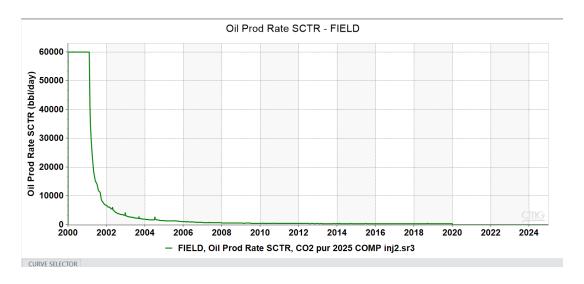


Figure 4.7. Oil production rate vs time for 100% CO_2 injection (generated by CMG Results, 2021).

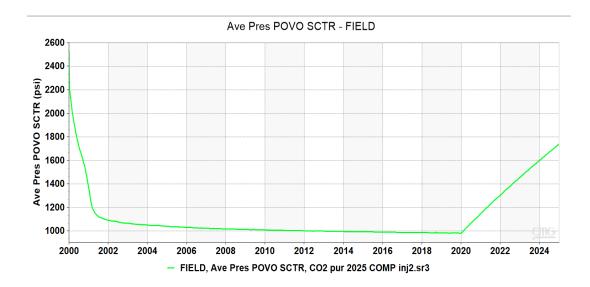


Figure 4.8. Average pressure povo SCTR vs time for 100% CO₂ injection (generated by CMG Results, 2021).

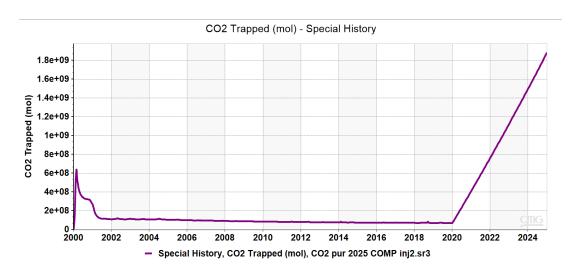


Figure 4.9. Cumulative CO₂ trapped vs time for 100% CO₂ injection (generated by CMG Results, 2021).

Case 3: Impure CO₂ Flooding

Carbon dioxide (CO_2) can be captured from a variety of origins using different technologies. However, these CO_2 streams often contain several impurities, including nitrogen (N_2) , methane (CH_4) , hydrogen sulfide (H_2S) , and argon (A_r) . The concentration of these impurities in CO_2 is influenced by the extent of fuel oxidation as well as the specific characteristics of the fuel utilized. Among the prevalent carbon capture techniques: pre-combustion, post-combustion, and oxy-fuel combustion; the

oxy-fuel combustion process results in the highest concentration of impurities within the CO_2 output.

In addition to these major contaminants, minor constituents like water vapor (H_2O) , nitrogen oxides (NO_X) , and sulfur oxides (SO_X) are also present. For this study, we will focus specifically on the effect of nitrogen (N_2) on the miscibility of CO_2 during enhanced oil recovery processes. The following table illustrates the minimum miscibility pressures (MMP) for various blends of CO_2 and N_2 , calculated using three different correlations: multiple mixing, cell to cell, and tie lines.

Table 4.2. MMP for impure CO₂ (generated by CMG Winprop, 2021).

Case	Cell to Cell	Tie Lines	Multiple Mixing
25% N ₂ + 75%	2700 psi	2210 psi	2900 psi
CO_2			
$50\% \ N_2 + 50\%$	2920 psi	2500 psi	3200 psi
CO_2			
$75\%\ N_2 + 25\%$	3080 psi	2820 psi	3600 psi
CO_2			

The data indicates that the miscibility characteristics are significantly influenced by the gas composition, particularly the presence of N₂, which elevates the minimum miscibility pressure required. As seen below, this scenario contains varying proportions of nitrogen and carbon dioxide:

• 25% $N_2 + 75\%$ CO_2 : An examination of a mixture comprising 25% N_2 and 75% CO_2 reveals that this injection scenario achieved an oil recovery rate of 69%, resulting in a cumulative production of 36.4 million barrels. This performance is notably lower than that observed in scenarios using 100% CO_2 injection. Below, we present the metrics for oil recovery factors, cumulative oil produced, and related production rates.

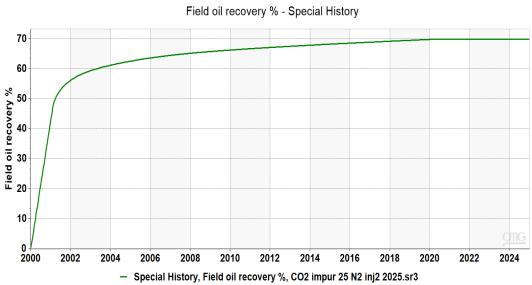


Figure 4.10. Oil recovery factor vs time for 25% $N_2 + 75\%$ CO₂ (generated by CMG Results, 2021).

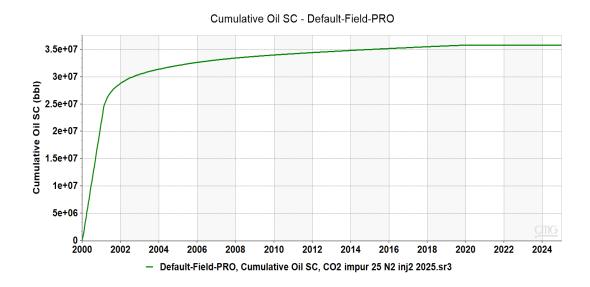


Figure 4.11. Cumulative oil production vs time for $25\% N_2 + 75\% CO_2$ (generated by CMG Results, 2021).

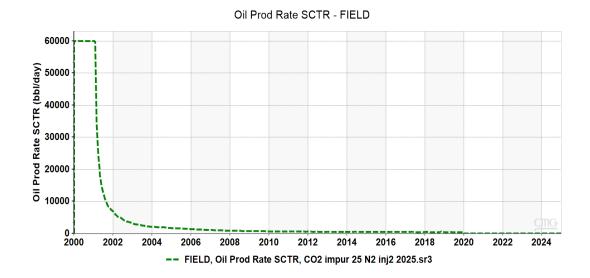


Figure 4.12. Oil production rate vs time for 25% $N_2 + 75\%$ CO₂ (generated by CMG Results, 2021).

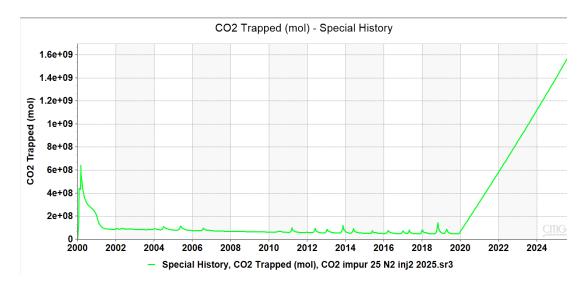


Figure 4.13. Cumulative CO_2 trapped vs time for 25% N_2 + 75% CO_2 (generated by CMG Results, 2021).

• 50% $N_2 + 50\%$ CO_2 : In contrast to the earlier injection scenario, the current injection process has resulted in a total oil production of 36.1 million barrels, achieving an oil recovery rate of 65%. The accompanying figures illustrate the rates of oil production, the cumulative amount of oil produced, and the recovery factor of the oil.

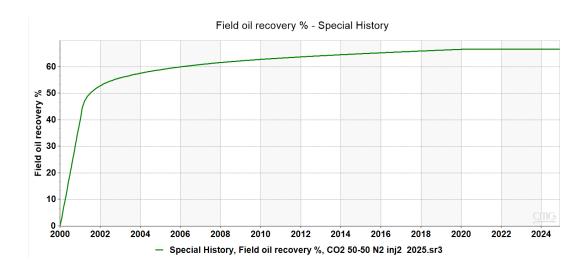


Figure 4.14 Oil recovery factor vs time for $50\%\,N_2+50\%\,CO_2$ (generated by CMG Results, 2021).

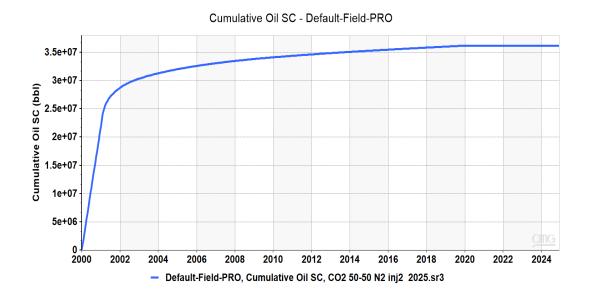


Figure 4.15. Cumulative oil production vs time for $50\% N_2 + 50\% CO_2$ (generated by CMG Results, 2021).

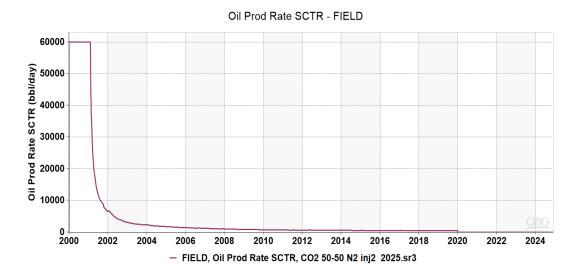


Figure 4.16. Oil production rate vs time for $50\% N_2 + 50\% CO_2$ (generated by CMG Results, 2021).

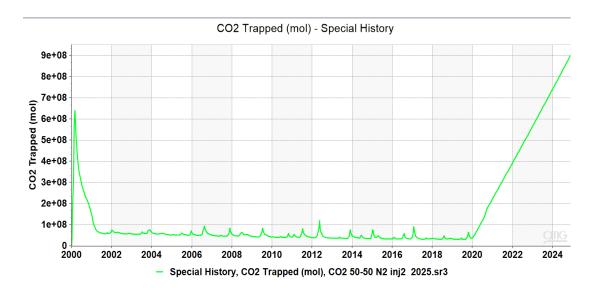


Figure 4.17. Cumulative CO_2 trapped vs time for 50% $N_2 + 50\%$ CO_2 (generated by CMG Results, 2021).

• $75\% N_2 + 25\% CO_2$: Compared to the earlier injection scenario, the recent injection has resulted in a cumulative production of 35.6 million barrels of oil, with an overall recovery rate of 65%. The figures below depict the oil production rate, the total cumulative oil produced, and the oil recovery factor.

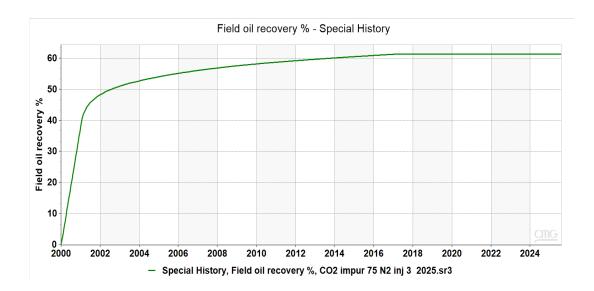


Figure 4.18. Oil recovery factor vs time for $75\% \, N_2 + 25\% \, CO_2$ (generated by CMG Results, 2021).

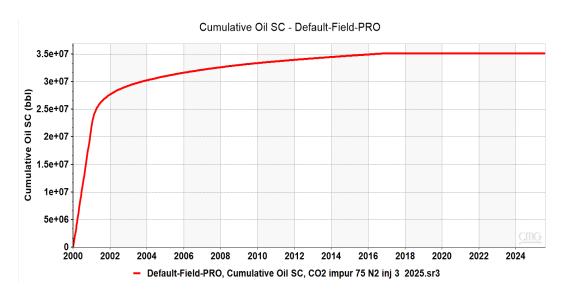


Figure 4.19. Cumulative oil vs time for $75\% N_2 + 25\% CO_2$ (generated by CMG Results, 2021).

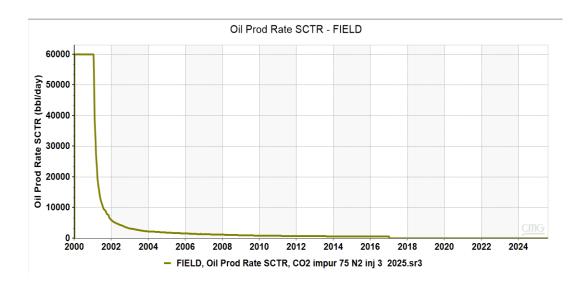


Figure 4.20. Oil production rate vs time for $75\% N_2 + 25\% CO_2$ (generated by CMG Results, 2021).

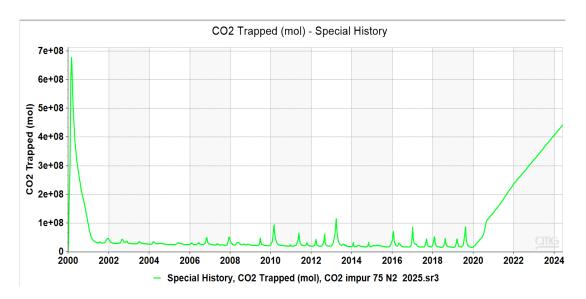


Figure 4.21. Cumulative CO_2 trapped vs time for $75\%N_2 + 25\%$ CO_2 (generated by CMG Results, 2021).

Case 4: 100% N₂

The following table will show the different values of pressure at FCM and MCM of the miscibility of carbon dioxide and nitrogen with oil using the following correlations: multiple mixing, cell to cell, and tie lines.

Table 4.3. MMP for pure N_2 (generated by CMG Winprop, 2021).

Case	Cell to Cell	Tie Lines	Multiple Mixing
100% N ₂	3150 psi	3000 psi	4187 psi

Compared to the previous injection scenario, the injection has yielded a reduced cumulative oil production of 35 million barrels and a 58% oil recovery. Figures presented below display the oil production rate, cumulative oil production, and oil recovery factor.

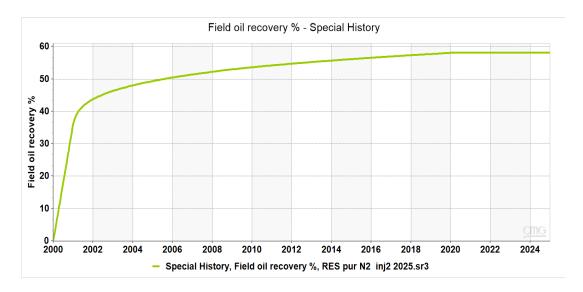
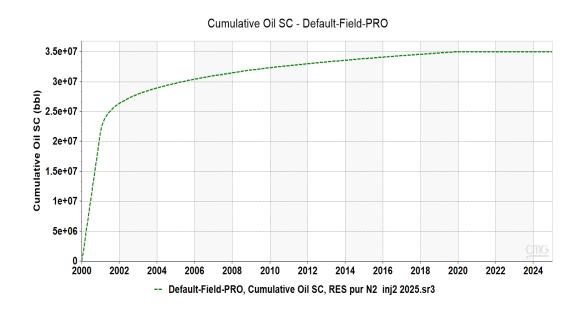


Figure 4.22. Oil recovery factor vs time for $100\% N_2$ (generated by CMG Results, 2021).



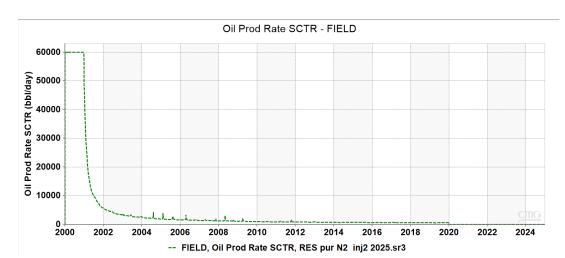


Figure 4.23. Cumulative oil vs time for 100% N₂ (generated by CMG Results, 2021).

Figure 4.24. Oil production rate vs time for 100% N₂ (generated by CMG Results, 2021).

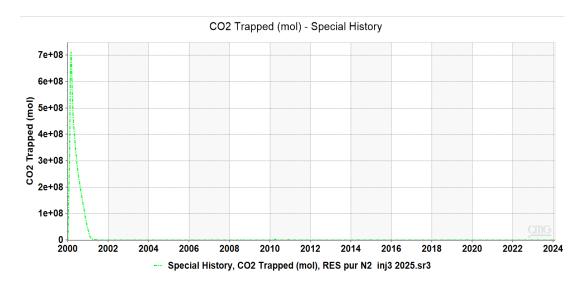


Figure 4.25. Cumulative mole trapped vs time for $100\% N_2$ (generated by CMG Results, 2021).

Comparison of Injection Scenarios With The Base Base

Oil Recovery: The main objective of this research is to enhance the oil recovery factor, quantified as the percentage of natural oil extracted from the original volume of oil in place (OOIP). Cumulative oil production refers to the total quantity of hydrocarbons recovered from the reservoir. Previous studies, including those by Khan et al. (2001, 2003), investigated a miscible CO₂ injection scenario, achieving an

oil recovery factor of 32% after 20 years. In contrast, the optimal scenario explored in this project, which entailed 25 years of miscible CO_2 injection, realized a significantly higher oil recovery rate of 73.8%, surpassing the outcomes noted in prior literature. The graphical data demonstrate that using pure miscible CO_2 yields the highest oil recovery factor when compared to CO_2 mixed with nitrogen (N_2) .

Table 4.4. Recovery factor and cumulative oil results of natural depletion, miscible pure CO_2 , impure CO_2 and pure N_2 .

Case	Natural	100%	25% N ₂ +	50% N ₂ +	75% N ₂ +	100% N ₂
	depletion	CO ₂	75%CO ₂	50%CO ₂	25%CO ₂	
Recovery factor%	27.3%	73.8%	69%	65%	62%	58%
Cumulative oil	10 MMbbl	38 MMbbl	36.4 MMbbl	36.1 MMbbl	35.6 MMbbl	35 MMbbl

The scenario involving the injection of pure CO_2 demonstrates a significant enhancement in both oil recovery and total cumulative oil production when compared to the baseline case. Over a period of 25 years, the pure CO_2 injection achieved an oil recovery factor of 73.8% and a cumulative oil production of 38 million barrels, as indicated in table 6. In contrast, the natural depletion process resulted in only 27.3% oil recovery with total cumulative production reaching 10 million barrels, also outlined in table 6.

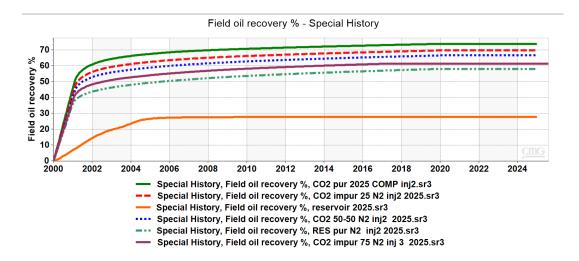


Figure 4.26. Comparison of field oil recovery values in natural depletion, pure CO_2 , impure CO_2 , and pure N_2 injection scenarios vs time (generated by CMG Results, 2021).

Cumulative Oil

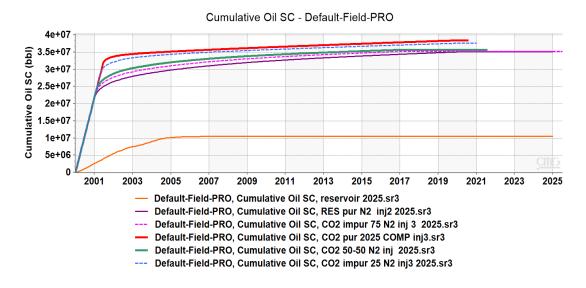


Figure 4.27. Comparison of cumulative oil values in natural depletion, pure CO_2 , impure CO_2 and pure N_2 injection scenarios vs time (generated by CMG Results, 2021).

Similarly, Figure 24 illustrates the comparison of cumulative oil production across these scenarios versus time. It is noteworthy that certain scenarios terminate in 2021 while others extend until 2025. This discrepancy is attributed to specific constraints applied to the producers in 2021. Production was halted while injection into the reservoir continued until the bottom hole pressure was achieved.

Production Rate

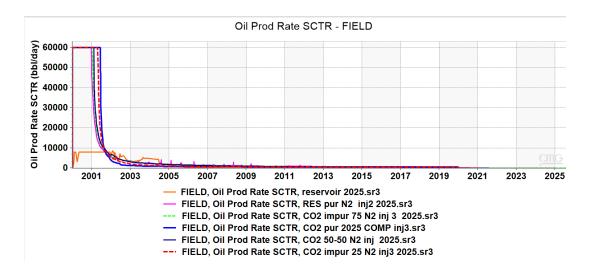


Figure 4.28. Comparison of production oil rate values in natural depletion, pure CO_2 , impure CO_2 , and pure N_2 injection scenarios vs time.

Average Reservoir Pressure: Average reservoir pressure serves as a significant indicator of the remaining fluid (gas, oil, or water) in a reservoir at any given moment. It reflects the potential force available to extract the remaining fluid during the simulation. The presence of N₂ occupies additional pore space and consequently increases the average reservoir pressure due to their lower compressibility compared to CO₂. However, in pure CO₂ scenarios, a greater flow pressure is required to maintain the same bottom hole pressure, as impurities decrease the density of the injected fluid. The ensuing figure demonstrates that contaminants diminish vertical sweep and displacement efficiency, leading to an increase in average reservoir pressure and a corresponding reduction in oil recovery. Specifically, pure nitrogen injection decrease oil recovery by 25.8% compared to pure carbon dioxide.

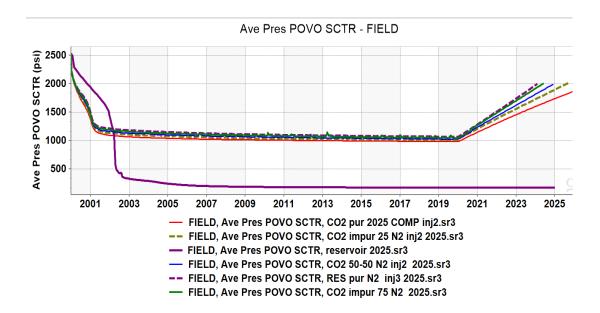


Figure 4.29. Comparison of average reservoir pressure values in natural depletion, pure CO_2 , impure CO_2 , and pure N_2 injection scenarios vs time (generated by CMG Results, 2021).

CO₂ Trapped

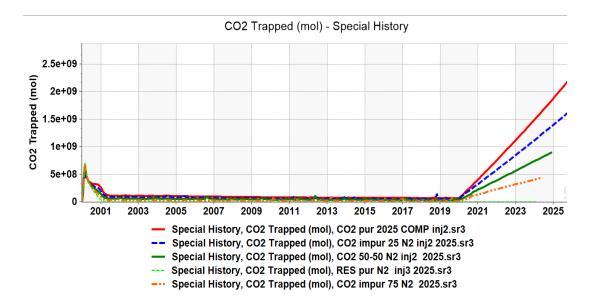


Figure 4.30. Comparison of CO_2 trapped for pure CO_2 , impure CO_2 , and pure N_2 injection scenarios vs time (generated by CMG Results, 2021).

Sweep Efficiency: The decline in sweep efficiency associated with contaminants in the CO_2 can be attributed to the viscosity of the oil. N_2 in the CO_2 stream hinder its ability to reduce the oil's viscosity, resulting in a faster migration of the injected gas. Consequently, the presence of N_2 leads to an increase in both viscosity

and interfacial tension of the oil, suggesting that reduced solvent capacity exists for the injected gas to lower the viscosity after multiple interactions. The subsequent figure reveals that an injection of $50\% N_2 + 50\% CO_2$ results in higher oil saturation and, consequently, decreased the displacement sweep efficiency inside the reservoir.

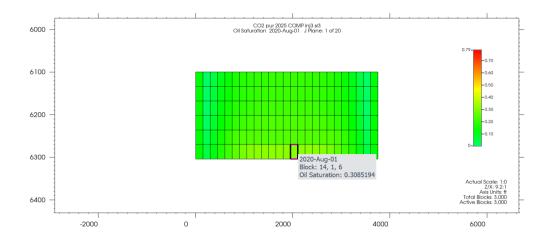


Figure 4.31. 2D aerial view of oil saturation for miscible pure CO₂ (generated by CMG Results, 2021).

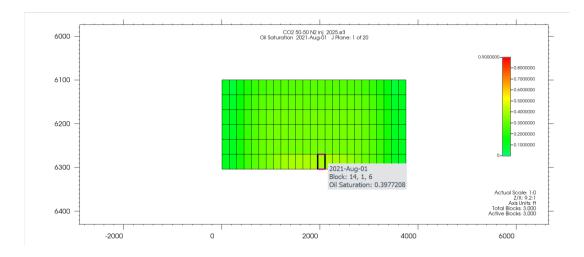


Figure 4.32. 2D aerial view of oil saturation for 50% N_2 + 50% CO_2 (generated by CMG Results, 2021).

MMP: The physical properties of gas can significantly influence both displacement and efficiency. In this study, the minimum miscibility pressure (MMP) of CO_2 with the reservoir fluid was determined using Winprop software and juxtaposed with the results obtained from the simulations. It is established that the purity level of CO_2 plays a crucial role in determining the MMP between oil and the injected solvent. Typically, the presence of nitrogen (N_2) within the CO_2 stream leads to an elevation

in the MMP. The outcomes from the MMP simulations with varying correlations of CO_2 and N_2 are presented in table 7. The inclusion of N_2 impurities correspondingly resulted in higher MMP values. Hence, any alterations in the gas's physical characteristics are likely to impact both the displacement efficiency and overall effectiveness of the process.

Table 4.5. MMP values from different correlations.

Case	Cell to Cell	Tie Lines	Multiple Mixing
Pure CO ₂	2575 psi	1950 psi	2623 psi
$75\% \text{ CO}_2 + 25\%$ N_2	2700 psi	2210 psi	2900 psi
50% CO ₂ + 50% N ₂	2920 psi	2500 psi	3200 psi
$25\% \text{ CO}_2 + 75\%$ N_2	3080 psi	2820 psi	3600 psi
100% N ₂	3150 psi	3000 psi	4187 psi

This table outlines an overview of three different techniques for multi-phase mixing in reservoirs. The multiple mixing cell method allows for an in-depth analysis of miscibility development through sequential contacts, making it ideal for detailed investigations where ample computational resources are accessible. On the other hand, the tie line method provides a simple and swift evaluation, although it may compromise accuracy when dealing with intricate mixtures. Lastly, the cell-to-cell simulation method portrays more realistic conditions within the reservoir but requires significant computational resources to implement effectively.

Environmental Impacts of CO₂-EOR

A lifecycle analysis of enhanced oil recovery indicates that CO₂ EOR approaches carbon neutrality, possesses a significant capacity for CO₂ storage, and

generates minimal process emissions. By mitigating risks associated with global warming, acidification, human toxicity, and eutrophication, as well as reducing waste and resource consumption, CO_2 -EOR presents substantial environmental benefits compared to traditional oil extraction methods. Notably, one of the key advantages of sequestering CO_2 in oil reservoirs is their considerable storage capacity. Our research has revealed that the Arab-D reservoir can store approximately 2.4 x 10^9 moles of CO_2 , which is a highly significant. Furthermore, oil reservoirs also have the ability to capture impurities found in carbon dioxide. The data illustrate that a significant quantity of nitrogen (N_2) is sequestered within these oil reservoir, which positively influences the environment, as nitrogen is classified as one of the greenhouse gases. This sequestration plays a role in lessening greenhouse gas emissions that can deteriorate the ozone layer, with about 6.4 accounted for. Since oil reservoirs have been effectively retaining oil for millions of years, they also serve as efficient "traps" for CO_2 and N_2 , offering a long operational lifespan that enhances safety considerations.

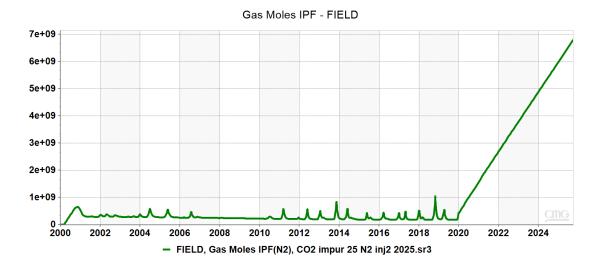


Figure 4.33. Cumulative N_2 trapped vs time for 25% N_2 + 75% CO_2 (generated by CMG Results, 2021).

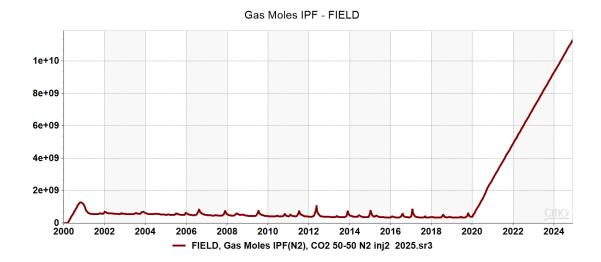


Figure 4.34. Cumulative N_2 trapped vs time for 50% N_2 + 50% CO_2 (generated by CMG Results, 2021).

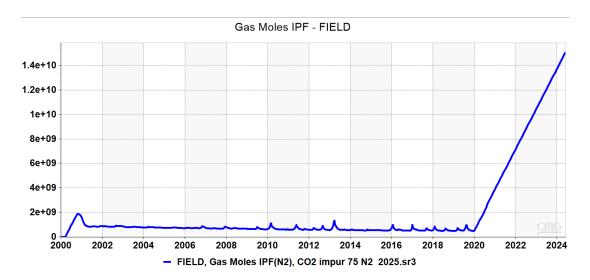


Figure 4.35. Cumulative $\rm N_2$ trapped vs time for 75% $\rm N_2$ + 25% $\rm CO_2$ (generated by CMG Results, 2021).

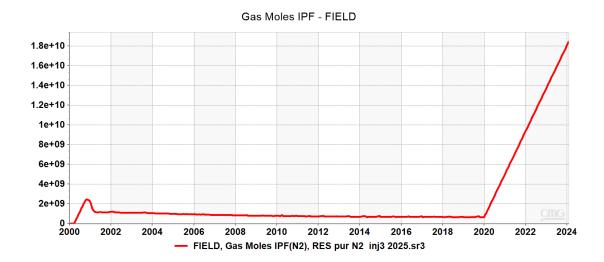


Figure 4.36 Cumulative N₂ trapped vs time for 100% N₂ (generated by CMG Results, 2021).

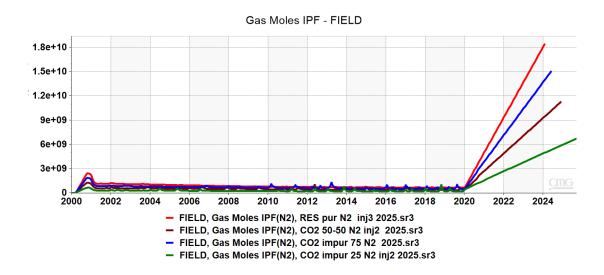


Figure 4.37. Comparison of N₂ trapped for injection scenarios vs time (generated by CMG Results, 2021).

The table below provides a detailed summary of the specific volumes of carbon dioxide and nitrogen that were effectively captured in the Arab-D reservoir during our research. This information not only demonstrates the high efficiency of our production rate but also highlights the reservoir's considerable storage capacity. Such capacity is essential as it plays a critical role in advancing environmental sustainability and supporting a healthier ecosystem. The subsequent table outlines the different volumes of carbon dioxide and nitrogen (CO_2/N_2) captured in the Arab-D reservoir throughout our modeling. This information showcases the notable production rate we achieved

while also emphasizing the reservoir's significant storage capability, which is instrumental in fostering a sustainable environment.

Table 4.6. Volume of CO_2/N_2 trapped for different scenarios.

Case	CO ₂ trapped(mol)	N ₂ trapped(mol)
Pure CO ₂	2.7*10°	0
75% CO ₂ + 25% N ₂	1.6*10 ⁹	6.8*10 ⁹
50% CO ₂ + 50% N ₂	9*10 ⁸	$1.1*10^{10}$
25% CO ₂ + 75% N ₂	4.4*108	1.5*10 ¹⁰
100% N ₂	0	$1.8*10^{10}$

CHAPTER V

Discussion

In reservoirs when natural energy is exhausted and only a tiny amount of economically viable oil can be produced, enhanced oil recovery is used. Injecting CO_2 helps to dissolve oil, reduce viscosity, and increase flow properties including surface area, wettability, etc. However, the miscibility and consequently the performance of the CO_2 are affected by the fluid composition, reservoir parameters, and pollutants present in the captured liquid. By storing CO_2 in subterranean reservoirs, EOR technology can help reduce rising carbon emissions and global warming. The findings of this study demonstrate that, in comparison to natural depletion, miscible flooding of pure CO_2 results in a very high oil recovery of 73.8%. Depending on the quantity of impurities in the CO_2 , the impure CO_2 flood in the alternative model receives a lower overall recovery, ranging from 65 to 62%. This is because the presence of contaminants in the injected solvent causes the flooding process to have a poor mobility ratio, which raises the oil's viscosity. Nitrogen reduces displacement sweep efficiency and raises oil viscosity in the EOR process, which explains why N_2 injection has a 58% oil recovery in contrast.

In terms of cumulative oil produced, which is another crucial element to take into account when assessing the overall effectiveness of an EOR operation, pure CO₂ injection generated 38 MMbbl of cumulative oil. In contrast, the injection of impure CO₂ resulted in a total oil production of 35.6–36.4 MMbbbl respectively in descending order of nitrogen injection throughout the simulated years, while the injection of N₂ pure produced an oil of 35 MMbbl. Because of the miscibility of the gas injected with the reservoir oil, the pure CO₂ performance clearly shows the increased cumulative oil produced. Lastly, the impure CO₂ flood had an oil displacement sweeping efficiency of roughly 40–50%. However, the removal of contaminants from CO₂ will result in a 70–75% increase in oil displacement sweeping efficiency. Finally, our results demonstrated that miscible pure CO₂ injection performed better than impure CO₂ injection.

In their research, Fong et al. (2016) investigated how various gas mixtures influence oil recovery, focusing on a blend comprising 80% nitrogen and 20% carbon

dioxide. Their results showed a significant reduction in recovery efficiency, with merely 23% of the oil initially in place (OOIP) being extracted, compared to a much higher 56% recovery rate achieved using pure carbon dioxide. This outcome aligns with the trends observed in our research, as the early breakthrough of gas in nitrogenrich mixtures underscores the considerable challenges and inefficiencies related to nitrogen's role in oil recovery processes.

In a similar vein, Spivak and Chima (2019) analyzed a mixture of 82% carbon dioxide and 18% nitrogen. Their findings revealed that this combination also resulted in lower recovery rates and earlier gas breakthroughs than those achieved with pure carbon dioxide. The limited ability of nitrogen to dissolve in Wilmington oil further compromised the overall efficiency of carbon dioxide solubility, leading to reduced performance which are therefore compatibles with the results of our investigation.

Additionally, Nguyen and Ali (1998) looked into the effects of replacing pure carbon dioxide with an impure version in the water-alternating-gas (WAG) method. Their study indicated that such a substitution could cause oil recovery losses of up to 10% due to a high concentration of nitrogen. In contrast, our findings show a loss of 25.8% when comparing pure nitrogen to pure carbon dioxide. This highlights the detrimental effects that elevated nitrogen levels and impurities within gas mixtures have on recovery efficiency, further demonstrating that scenarios utilizing pure carbon dioxide yield significantly improved results in oil extraction processes.

In summary, the collective evidence from these studies illustrates that the incorporation of nitrogen and other impurities in gas mixtures consistently leads to reduced oil recovery rates and operational difficulties when compared to scenarios involving pure carbon dioxide.

To produce high purity CO_2 , impurities collected with it undergo a purification step, which raises the capture cost considerably. Injecting the CO_2 and impurities together is hence economical. On the other hand, impure CO_2 has a negative impact on CO_2 -EOR performance, which leads to less oil recovery. The variation in oil recovery based on the degree of purity will be managed and taken into account. Even though it was not taken into account in this study, it is crucial to create an economic model that uses economic factors like capital expenditures (CAPEX), operating expenditures (OPEX), tax credits, oil price, and CO_2 price depending on the purity

level, sources, and technology in order to estimate the proper purity level of the $\rm CO_2$ stream. Field applications need the development of a design for each site, including various injection scenarios, based on economic analysis because the best operating conditions to boost CCS-EOR performance rely on the reservoir's features.

CHAPTER VI

Conclusion and Recommendations

This thesis's numerous investigations have illustrated how various parameters affect the use of miscible $\mathbf{CO_2}$ flooding as an EOR method in addition to other CCS strategies. The primary findings of this study are presented in this chapter. Furthermore, some suggestions are made for further research so that the scientists can better understand how to predict the injection of $\mathbf{CO_2}$ miscible gas to enhance oil recovery.

Conclusion

For the purpose of addressing the world's energy shortfall, the petroleum and natural gas industries' adoption of improved oil recovery has greatly boosted the production of hydrocarbons. Injecting chemicals to enhance oil recovery or storing CO_2 underground to lower carbon emissions into the atmosphere are only two of the methods. The purpose of the simulation was to predict the oil recovery, production rates, and pressure in the Ghawar oil field under miscible gas injection. A model may be constructed or run many times at low cost in a short amount of time, but the field can only be produced once at high cost. In order to draw a conclusion following successful simulations, it was determined that miscible pure CO_2 injection has significantly increased oil recovery based on the various scenarios used for this project, ranging from the no injection scenario to the N_2 flooding scenarios. As a result, this strengthens the reservoir and delays its early depletion.

- ➤ In addition to the oil recovery of 73.8% and cumulative oil production of 38 MMbbl, the pure miscible CO₂ model appears to be substantially larger than the basic model and the various injections of impure CO₂.
- ➤ In contrast to the 73.8% OOIP recovery attained during our investigation, a study using 100% pure carbon dioxide revealed that 56% OOIP was recovered (Fong et al., 2000).
- At the same injection pressure of 3212 psi, the gas with the highest oil recovery factor was CO₂, whereas the gas with the lowest oil recovery factor was N₂.

Furthermore, compared to CO_2 , which had the longest breakthrough time and the lowest MMP, N_2 had the quickest breakthrough time and the highest MMP. This is because CO_2 has a greater absorptive capacity and is hence condensable, whereas N_2 has a very low adsorptive capacity and is consequently a noncondensable gas.

The injection of pure nitrogen lowered the overall CCS performance by 38% compared to pure carbon dioxide injection. The Arab-D reservoir has also caught $2.7*10^9$ mol of CO_2 and $1.8*10^{10}$ mol of N_2 corresponding respectively to $1.1*10^5$ tonnes and $5.04*10^5$ tonnes, in addition to producing a very high oil recovery after the injection of pure CO_2 .

Recommendations

Recommendations According to Findings

- Additionally, a global opportunity for carbon capture and sequestration (CCS) is offered by the injection of miscible gas (CO₂). This CCS approach can be further modeled by adopting and improving the model.
- Adequate field data and laboratory work are necessary to effectively model EOR projects using CMG software, particularly as it entails the use of solvents and other chemicals and nanomaterials.

Recommendations for Further Research

➤ Even though miscible pure CO₂ injection produced the highest oil recovery, it is still advantageous to do an economic analysis to determine the cost per barrel of purification in order to assess the investment's viability. Future studies may focus on this topic.

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Appendices

Appendix A

Modeling Miscible $\mathrm{CO}_2/\mathrm{N}_2$ Injection for Enhanced Oil Recovery CMG GEM Data

INUNII FIELD
WSRF WELL 1
WSRF GRID TIME
OUTSRF GRID PRES SG SIG SO SW VISO
OUTSRF RES NONE
OUTSRF WELL GHGSOL
GHGTHY
PAVG
RECG
RECO
WPRN GRID 0
OUTPRN GRID NONE
OUTPRN RES NONE
Definition of fundamental cartesian grid
GRID VARI 25 20 6
KDIR DOWN
DI IVAR
25*150
DJ JVAR
20*150
DK ALL
3000*34
DTOP
500*6100

XOFFSET 0.0000

YOFFSET 0.0000

ROTATION 0.0000

AXES-DIRECTIONS 1.0 -1.0 1.0

NULL CON 1

POR CON 0.19

PERMI CON 617

PERMJ CON 617

PERMK CON 61

PINCHOUTARRAY CON 1

PRPOR 2500

CPOR 0.000004

The following is the fluid component

property data in GEM format.

The unit system and fluid compositions should

be specified in the I/O control section.

The units and compositions specified in WinProp

are included here as comments for informational purposes.

PVT UNITS CONSISTENT WITH *INUNIT *FIELD

COMPOSITION PRIMARY

0.01590000 0.34280000 0.07520000 0.05640000 0.00970000

 $0.02490000 \ 0.01100000 \ 0.01410000 \ 0.01970000 \ 0.43030000$

COMPOSITION *SECOND

Model and number of components

MODEL PR

NC 10 10

COMPNAME 'CO2' 'CH4' 'C2H6' 'C3H8' 'IC4' 'NC4' 'IC5' 'NC5' 'FC6' 'C7+'

The Hydrocarbon component flag values: 1 - Hydro-Carbon, 0 - non Hydro-Carbon

The read-in HC-HC BIN values will be overwritten by the internal GEM calculated

values for these HC-HC pairs, if the original *HCFLAG values are used.

This is fine if there is only 1 HC-HC group defined in the original WinProp data set. Otherwise, all-zero *HCFLAG will be output here in order to use the full read-in BIN values.

The original WinProp HC-flags:

HCFLAG 3 1 1 1 1 1 1 1 1 1

HCFLAG

0111111111

TRES 215.00000

PVC3 1.20000000E+00

VISCOR HZYT

MIXVC 1.0000000E+00

VISCOEFF

1.0230000E-01 2.3364000E-02 5.8533000E-02 -4.0758000E-02 9.3324000E-03

MW

4.4010000E+01 1.6043000E+01 3.0070000E+01 4.4097000E+01 5.8124000E+01 5.8124000E+01 7.2151000E+01 7.2151000E+01 8.6000000E+01 2.1500000E+02

AC

2.2500000E-01 8.0000000E-03 9.8000000E-02 1.5200000E-01 1.7600000E-01 1.9300000E-01 2.2700000E-01 2.5100000E-01 2.7504000E-01 6.6096582E-01

PCRIT

7.2800000E+01 4.5400000E+01 4.8200000E+01 4.1900000E+01 3.6000000E+01 3.7500000E+01 3.3400000E+01 3.3300000E+01 3.2460000E+01 1.5284295E+01

VCRIT

9.4000000E-02 9.9000000E-02 1.4800000E-01 2.0300000E-01 2.6300000E-01 2.5500000E-01 3.0600000E-01 3.0400000E-01 3.4400000E-01 8.9560826E-01

TCRIT

3.0420000E+02 1.9060000E+02 3.0540000E+02 3.6980000E+02 4.0810000E+02 4.2520000E+02 4.6040000E+02 4.6960000E+02 5.0750000E+02 7.2382387E+02

PCHOR

7.8000000E+01 7.7000000E+01 1.0800000E+02 1.5030000E+02 1.8150000E+02 1.8990000E+02 2.2500000E+02 2.3150000E+02 2.5010880E+02 5.8135500E+02

SG

8.1800000E-01 3.0000000E-01 3.5600000E-01 5.0700000E-01 5.6300000E-01 5.8400000E-01 6.2500000E-01 6.3100000E-01 6.9000000E-01 8.0000000E-01

TB

-1.0921000E+02 -2.5861000E+02 -1.2757000E+02 -4.3690000E+01 1.0670000E+01 3.1190000E+01 8.2130000E+01 9.6890000E+01 1.4693000E+02 5.3421523E+02

OMEGA

4.5723553E-01 4.5723553E-01

OMEGB

7.7796074E-02 7.7796074E-02

VSHIFT

HEATING_VALUES

0.0000000E+00 8.4429001E+02 1.4784600E+03 2.1051600E+03 2.7115400E+03 2.7115400E+03 3.3536600E+03 3.3536600E+03 3.9759100E+03 0.0000000E+00

VISVC

9.4000000E-02 9.9000000E-02 1.4800000E-01 2.0300000E-01 2.6300000E-01 2.5500000E-01 3.0600000E-01 3.0400000E-01 3.4400000E-01 8.9560826E-01

BIN

1.0500000E-01

- 1.3000000E-01 2.6890022E-03
- 1.2500000E-01 8.5370405E-03 1.6620489E-03
- 1.2000000E-01 1.5715316E-02 5.4857876E-03 1.1165976E-03
- 1.1500000E-01 1.4748531E-02 4.9143360E-03 8.6625350E-04 1.5903506E-05
- 1.1500000E-01 2.0878892E-02 8.7338646E-03 2.8007353E-03 3.8207590E-04 5.5378054E-04
- 1.1500000E-01 2.0640839E-02 8.5779330E-03 2.7121325E-03 3.4971119E-04 5.1467786E-04 7.1665797E-07
- 1.1500000E-01 2.5345101E-02 1.1747825E-02 4.6198099E-03 1.2003051E-03 1.4920539E-03 2.2833006E-04 2.5462307E-04
- 0.0000000E+00 7.6042151E-02 5.1834174E-02 3.5697700E-02 2.4546345E-02 2.5774365E-02 1.8938276E-02 1.9166794E-02 1.5080614E-02

ENTHCOEF

- 9.6880000E-02 1.5884300E-01 -3.3712000E-05 1.4810500E-07 -9.6620300E-11 2.0738320E-14
- -2.8385700E+00 5.3828500E-01 -2.1140900E-04 3.3927600E-07 -1.1643220E-10 1.3896120E-14
- -1.4220000E-02 2.6461200E-01 -2.4568000E-05 2.9140200E-07 -1.2810330E-10 1.8134820E-14
- 6.8715000E-01 1.6030400E-01 1.2608400E-04 1.8143000E-07 -9.1891300E-11 1.3548500E-14
- 1.4595600E+00 9.9070000E-02 2.3873600E-04 9.1593000E-08 -5.9405000E-11 9.0964500E-15
- 7.2281400E+00 9.9687000E-02 2.6654800E-04 5.4073000E-08 -4.2926900E-11 6.6958000E-15
- 1.7694120E+01 1.5946000E-02 3.8244900E-04 -2.7557000E-08 -1.4303500E-11 2.9567700E-15
- 9.0420900E+00 1.1182900E-01 2.2851500E-04 8.6331000E-08 -5.4464900E-11 8.1845000E-15
- 0.0000000E+00 -1.6543463E-02 4.1169069E-04 -5.7742757E-08 0.0000000E+00 0.0000000E+00

0.0000000E+00 2.2135278E-03 3.9756979E-04 -5.3302731E-08 0.0000000E+00 0.0000000E+00

CW 0.000004

REFPW 2500

SOLUBILITY HENRY

TRACE-COMP 2

** HENRYC calculated at 215.00 deg F

HENRYC

8.4206782E+04 0 1.2941741E+06 1.8308289E+06 2.1086164E+06 2.1088205E+06 4.8095508E+06 4.8677661E+06 1.4695949E+21 1.4695949E+21

REFPH

2.1960000E+03 2.19600000E+03 2.19600000E+03 2.19600000E+03 2.19600000E+03 2.19600000E+

VINFINITY

3.5668181E-02 0 5.2487562E-02 7.1896258E-02 9.1466009E-02 9.1485931E-02 1.1055231E-01 1.1302877E-01 1.2497583E-01 3.7227583E-01

HENRY-MOD1-CO2

BIN-TDEP-CO2

AQUEOUS-DENSITY *LINEAR

ROCKFLUID

HYS_MODEL LINEAR

RPT 1

Sw krw krow

SWT

0.21 0 0.8

0.225 0.00117187 0.703125

0.25 0.0046875 0.6125

0.275 0.0105469 0.528125

0.3 0.01875 0.45

0.325 0.0292969 0.378125

0.35 0.0421875 0.3125

0.375 0.0574219 0.253125

0.4 0.075 0.2

0.425 0.0949219 0.153125

0.45 0.117187 0.1125

0.475 0.141797 0.078125

0.5 0.16875 0.05

0.525 0.198047 0.028125

0.55 0.229688 0.0125

0.575 0.263672 0.003125

0.6 0.3 0

** SI krg krog

SLT

0.4 0.3 0

0.434375 0.263672 0.003125

0.46875 0.229688 0.0125

0.503125 0.198047 0.028125

0.5375 0.16875 0.05

0.571875 0.141797 0.078125

0.60625 0.117188 0.1125

0.640625 0.0949219 0.153125

0.675 0.075 0.2

0.709375 0.0574219 0.253125

0.74375 0.0421875 0.3125

0.778125 0.0292969 0.378125

0.8125 0.01875 0.45 0.846875 0.0105469 0.528125 0.88125 0.0046875 0.6125 0.915625 0.00117188 0.703125 0.95 0 0.8 **HYSKRG** INITIAL VERTICAL BLOCK_CENTER WATER_OIL_GAS ZOIL 0.0159 0.3428 0.0752 0.0564 0.0097 0.0249 0.011 0.0141 0.0197 0.4303 **ZGAS** 0 0.3 0.07 0.06 0.0093 0.02 0.011 0.01 0.0197 0.5 **REFPRES** 2500 **REFDEPTH** 6100 **DWOC** 6600 **DGOC** 6000 **NUMERICAL** DTMAX 1000000 DTMIN 0.000001 NORM UNKNOWN **MAXCHANGE TEMP 15 CONVERGE PRESS 0.01**

RUN

DATE 2000 1 1

WELL 'inj'

INJECTOR 'inj'

OPERATE MAX STG 10000000.0 CONT REPEAT

OPERATE MAX BHP 2000.0 CONT REPEAT

rad geofac wfrac skin

GEOMETRY K 0.25 0.37 1.0 0.0

PERF GEOA 'inj'

UBA ff Status Connection

13 10 1 1.0 OPEN FLOW-FROM 'SURFACE' REFLAYER

13 10 2 1.0 OPEN FLOW-FROM 1

13 10 3 1.0 OPEN FLOW-FROM 2

13 10 4 1.0 OPEN FLOW-FROM 3

13 10 5 1.0 OPEN FLOW-FROM 4

13 10 6 1.0 OPEN FLOW-FROM 5

LAYERXYZ 'inj'

perf geometric data: UBA, block entry(x,y,z) block exit(x,y,z), length

13 10 1 1875.000000 1425.000000 6100.000000 1875.000000 1425.000000 6134.000000 34.000000

13 10 2 1875.000000 1425.000000 6134.000000 1875.000000 1425.000000 6168.000000 34.000000

13 10 3 1875.000000 1425.000000 6168.000000 1875.000000 1425.000000 6202.000000 34.000000

13 10 4 1875.000000 1425.000000 6202.000000 1875.000000 1425.000000 6236.000000 34.000000

13 10 5 1875.000000 1425.000000 6236.000000 1875.000000 1425.000000 6270.000000 34.000000

13 10 6 1875.000000 1425.000000 6270.000000 1875.000000 1425.000000 6304.000000 34.000000

OPEN 'inj'

WELL 'prod'

PRODUCER 'prod'

OPERATE MAX STO 15000.0 CONT

OPERATE MIN BHP 28.0 CONT

rad geofac wfrac skin

GEOMETRY K 0.25 0.37 1.0 0.0

PERF GEOA 'prod'

UBA ff Status Connection

2 19 1 1.0 OPEN FLOW-TO 'SURFACE' REFLAYER

2 19 2 1.0 OPEN FLOW-TO 1

2 19 3 1.0 OPEN FLOW-TO 2

2 19 4 1.0 OPEN FLOW-TO 3

2 19 5 1.0 OPEN FLOW-TO 4

2 19 6 1.0 OPEN FLOW-TO 5

LAYERXYZ 'prod'

perf geometric data: UBA, block entry(x,y,z) block exit(x,y,z), length

2 19 1 225.000000 2775.000000 6100.000000 225.000000 2775.000000 6134.000000 34.000000

2 19 2 225.000000 2775.000000 6134.000000 225.000000 2775.000000 6168.000000 34.000000

2 19 3 225.000000 2775.000000 6168.000000 225.000000 2775.000000 6202.000000 34.000000

2 19 4 225.000000 2775.000000 6202.000000 225.000000 2775.000000 6236.000000 34.000000

2 19 5 225.000000 2775.000000 6236.000000 225.000000 2775.000000 6270.000000 34.000000

2 19 6 225.000000 2775.000000 6270.000000 225.000000 2775.000000 6304.000000 34.000000

OPEN 'prod'

WELL 'prod2'

PRODUCER 'prod2'

OPERATE MAX STO 15000.0 CONT

OPERATE MIN BHP 28.0 CONT

rad geofac wfrac skin

GEOMETRY K 0.25 0.37 1.0 0.0

PERF GEOA 'prod2'

UBA ff Status Connection

24 2 1 1.0 OPEN FLOW-TO 'SURFACE' REFLAYER

24 2 2 1.0 OPEN FLOW-TO 1

24 2 3 1.0 OPEN FLOW-TO 2

24 2 4 1.0 OPEN FLOW-TO 3

24 2 5 1.0 OPEN FLOW-TO 4

24 2 6 1.0 OPEN FLOW-TO 5

LAYERXYZ 'prod2'

perf geometric data: UBA, block entry(x,y,z) block exit(x,y,z), length

24 2 1 3525.000000 225.000000 6100.000000 3525.000000 225.000000 6134.000000 34.000000

24 2 2 3525.000000 225.000000 6134.000000 3525.000000 225.000000 6168.000000 34.000000

24 2 3 3525.000000 225.000000 6168.000000 3525.000000 225.000000 6202.000000 34.000000

24 2 4 3525.000000 225.000000 6202.000000 3525.000000 225.000000 6236.000000 34.000000

24 2 5 3525.000000 225.000000 6236.000000 3525.000000 225.000000 6270.000000 34.000000

24 2 6 3525.000000 225.000000 6270.000000 3525.000000 225.000000 6304.000000 34.000000

OPEN 'prod2'

WELL 'PROD3'

PRODUCER 'PROD3'

OPERATE MAX STO 15000.0 CONT

OPERATE MIN BHP 28.0 CONT

rad geofac wfrac skin

GEOMETRY K 0.25 0.37 1.0 0.0

PERF GEOA 'PROD3'

UBA ff Status Connection

2 2 1 1.0 OPEN FLOW-TO 'SURFACE' REFLAYER

2 2 2 1.0 OPEN FLOW-TO 1

2 2 3 1.0 OPEN FLOW-TO 2

2 2 4 1.0 OPEN FLOW-TO 3

2 2 5 1.0 OPEN FLOW-TO 4

2 2 6 1.0 OPEN FLOW-TO 5

LAYERXYZ 'PROD3'

perf geometric data: UBA, block entry(x,y,z) block exit(x,y,z), length

- 2 2 1 225.000000 225.000000 6100.000000 225.000000 225.000000 6134.000000 34.000000
- 2 2 2 225.000000 225.000000 6134.000000 225.000000 225.000000 6168.000000 34.000000
- 2 2 3 225.000000 225.000000 6168.000000 225.000000 225.000000 6202.000000 34.000000
- 2 2 4 225.000000 225.000000 6202.000000 225.000000 225.000000 6236.000000 34.000000
- 2 2 5 225.000000 225.000000 6236.000000 225.000000 225.000000 6270.000000 34.000000
- 2 2 6 225.000000 225.000000 6270.000000 225.000000 225.000000 6304.000000 34.000000

OPEN 'PROD3'

WELL 'PROD4'

PRODUCER 'PROD4'

OPERATE MAX STO 15000.0 CONT

OPERATE MIN BHP 28.0 CONT

rad geofac wfrac skin

GEOMETRY K 0.25 0.37 1.0 0.0

PERF GEOA 'PROD4'

UBA ff Status Connection

24 19 1 1.0 OPEN FLOW-TO 'SURFACE' REFLAYER

24 19 2 1.0 OPEN FLOW-TO 1

24 19 3 1.0 OPEN FLOW-TO 2

24 19 4 1.0 OPEN FLOW-TO 3

24 19 5 1.0 OPEN FLOW-TO 4

24 19 6 1.0 OPEN FLOW-TO 5

LAYERXYZ 'PROD4'

perf geometric data: UBA, block entry(x,y,z) block exit(x,y,z), length

24 19 1 3525.000000 2775.000000 6100.000000 3525.000000 2775.000000 6134.000000 34.000000

24 19 2 3525.000000 2775.000000 6134.000000 3525.000000 2775.000000 6168.000000 34.000000

24 19 3 3525.000000 2775.000000 6168.000000 3525.000000 2775.000000 6202.000000 34.000000

24 19 4 3525.000000 2775.000000 6202.000000 3525.000000 2775.000000 6236.000000 34.000000

24 19 5 3525.000000 2775.000000 6236.000000 3525.000000 2775.000000 6270.000000 34.000000

24 19 6 3525.000000 2775.000000 6270.000000 3525.000000 2775.000000 6304.000000 34.000000

OPEN 'PROD4'

GUIDEI STG 'inj'

1e+07

*MXCNRPT 100

DATE 2000 2 1.00000

DATE 2000 3 1.00000

DATE 2000 4 1.00000

DATE 2000 5 1.00000

DATE 2000 6 1.00000

DATE 2000 7 1.00000

DATE 2000 8 1.00000

DATE 2000 9 1.00000

DATE 2000 10 1.00000

DATE 2000 11 1.00000

DATE 2000 12 1.00000

DATE 2001 1 1.00000

DATE 2001 2 1.00000

DATE 2001 3 1.00000

DATE 2001 4 1.00000

DATE 2001 5 1.00000

DATE 2001 6 1.00000

DATE 2001 7 1.00000

DATE 2001 8 1.00000

DATE 2001 9 1.00000

DATE 2001 10 1.00000

DATE 2001 11 1.00000

DATE 2001 12 1.00000

DATE 2002 1 1.00000

DATE 2002 2 1.00000

DATE 2002 3 1.00000

DATE 2002 4 1.00000

DATE 2002 5 1.00000

DATE 2002 6 1.00000

DATE 2002 7 1.00000

DATE 2002 8 1.00000

DATE 2002 9 1.00000

DATE 2002 10 1.00000

DATE 2002 11 1.00000

DATE 2002 12 1.00000

DATE 2003 1 1.00000

DATE 2003 2 1.00000

DATE 2003 3 1.00000

DATE 2003 4 1.00000

DATE 2003 5 1.00000

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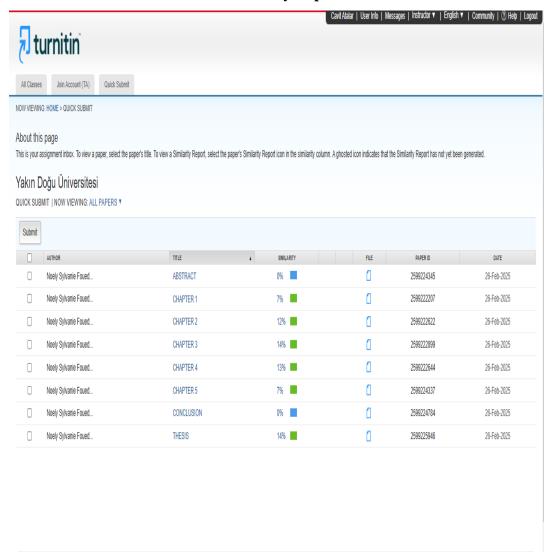
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Appendix B Similarity Report



APPENDIX C

ETHICAL APPROVAL LETTER



YAKIN DOĞU ÜNİVERSİTESİ

ETHICAL APROVAL DOCUMENT

Date: 10/04/2025

To the Institute of Graduate Studies

The research project titled "MODELING MISCIBLE CO₂ INJECTION FOR ENHANCED OIL RECOVERY: AIN DAR AREA" has been evaluated. Since the researcher will not collect primary data from humans, animals, plants or earth, this project does not need through the ethics committee.

Title: Prof. Dr.

Name Surname: Cavit ATALAR

Signature:

Role in the Research Project: Supervisor