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ARTICLE

An Integrated Optimization Method for $CO₂$ Pre-Injection during Hydraulic Fracturing in Heavy Oil Reservoirs

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ABSTRACT

 $CO₂$ pre-injection during hydraulic fracturing is an important method for the development of medium to deep heavy oil reservoirs. It reduces the interfacial tension and viscosity of crude oil, enhances its flowability, maintains reservoir pressure, and increases reservoir drainage capacity. Taking the Badaowan Formation as an example, in this study a detailed three-dimensional geomechanical model based on static data from well logging interpretations is elaborated, which can take into account both vertical and horizontal geological variations and mechanical characteristics. A comprehensive analysis of the impact of key construction parameters on $Pre\text{-}CO_2$ based fracturing (such as cluster spacing and injection volume), is therefore conducted. Thereafter, using optimized construction parameters, a non-structured grid for dynamic development prediction is introduced, and the capacity variations of different production scenarios are assessed. On the basis of the simulation results, reasonable fracturing parameters are finally determined, including cluster spacing, fracturing fluid volume, proppant concentration, and well spacing.

KEYWORDS

Heavy oil reservoir; pre-storage $CO₂$ energy fracturing; horizontal well fracturing parameters; numerical simulation

Nomenclature

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Greek Symbols

1 Introduction

Vast unconventional oil and gas reserves exist globally, with heavy oil reservoirs being particularly significant as a promising alternative energy source for the future [[1](#page-18-0)[,2\]](#page-18-1). However, these unconventional reservoirs typically exhibit low permeability and porosity, high crude oil viscosity, and substantial heat loss during heavy oil thermal recovery, necessitating alternative methods to develop this type of reservoir effectively $[3,4]$ $[3,4]$ $[3,4]$. CO₂ has long been widely applied in oilfields $[5]$ $[5]$ $[5]$. On the one hand, compared to other non-hydrocarbon media, $CO₂$ exhibits better miscibility with crude oil. When dissolved in crude oil, it effectively reduces the interfacial tension and viscosity, enhancing the flow properties of crude oil in the reservoir. Therefore, for heavy oil with poorer flowability, higher viscosity, and greater surface tension, the use of $CO₂$ to enhance recovery is more crucial. On the other hand, the application of $CO₂$ resources in oilfield production development is a significant response to the carbon capture, utilization, and storage (CCUS) policy to mitigate the global climate crisis.

Some scholars $[6,7]$ $[6,7]$ proposed the application of supercritical $CO₂$ fluids in oil and gas drilling and fracturing. They proposed that using supercritical $CO₂$ fluids for shale reservoir fracturing can avoid a series of issues caused by traditional fracturing fluids, such as water resource utilization, liquid backflow, and waste generation, and conducted theoretical and experimental research, demonstrating its technical feasibility [[8](#page-18-7)]. Subsequently, universities and research institutions in many countries conducted a significant amount of basic research on the application of supercritical $CO₂$ in petroleum exploration and development [[9](#page-18-8)–[12\]](#page-19-0).

The pre-stored CO_2 fracturing technology utilizes liquid CO_2 as a medium, injecting a sufficient amount of CO2 into the reservoir, creating a complex network of fractures near the wellbore to improve reservoir drainage capacity. Meanwhile, under high-temperature and high-pressure conditions in the middle and deep reservoirs, CO₂ becomes supercritical and dissolves in the crude oil, resulting in volume expansion and viscosity reduction. In unconventional reservoirs, factors such as stress state, stress anisotropy, rock mechanical properties, and natural fractures can determine the growth of hydraulic fractures and significantly affect well productivity [\[13](#page-19-1)–[15\]](#page-19-2). In addition, reservoir physical parameters such as saturation, porosity, matrix permeability, etc., also affect production $[16,17]$ $[16,17]$ $[16,17]$. Therefore, it is important to simulate fracture propagation and evaluate the subsequent productivity.

This study proposes a workflow for the optimization design of pre- $CO₂$ storage hydraulic fracturing for medium and deep heavy oil reservoirs in the Badaowan Formation (Junggar Basin, China). The workflow, implemented by the Petrel integrated platform, holds significant importance for optimizing well spacing and designing of hydraulic fracturing operations in the medium and deep heavy oil reservoirs. In this study, based on the geological understanding of the Badaowan Formation (Junggar Basin, China), a workflow for the optimization design of pre-stored $CO₂$ fracturing in middle and deep heavy oil reservoirs was proposed using the Petrel integrated platform, which is of great significance for optimizing well spacing and reasonable design of fracturing operation. The workflow includes geological modeling, geomechanical modeling, fracturing simulation, and production simulation. Firstly, a geomechanical model using well log curves is established and the model reliability in comparison with field data is validated. Subsequently, hydraulic fracturing simulations utilizing the geomechanical model are conducted. Finally, an unstructured grid is used to discretize the fractures for reservoir numerical simulations [[18\]](#page-19-5). Based on the simulation results, the optimal values for fracturing parameters (cluster spacing, fracturing fluid volume, proppant concentration, well spacing, etc.) are determined and applied into the field.

2 Study Area

The studied area is located in the Wuxia Fault Belt at the northwest margin of the Junggar Basin. It is approximately 35 kilometers east of the Urho District in Karamay City. The surface terrain is characterized by desert and desert steppe, with an average elevation of 490.5 m above sea level. The Jurassic Badaowan Formation in the studies area has a calculated petroleum geological reserve of 10.70 million tons, covering an oil-bearing area of 7.95 square kilometers. The natural gas geological reserves are 53 million cubic meters, covering a gas-bearing area of 0.54 square kilometers. The formation thickness ranges from 16 to 52 meters, with an average thickness of about 36 meters. The predominant lithology of the formation includes medium sandstone, fine sandstone, followed by coarse sandstone and pebbly sandstone. Specifically, natural fractures are not well developed. Currently, there is one horizontal well (Well A) in the area, and the operators plan to drill two additional horizontal wells (Well B and Well C). Therefore, it is necessary to enhance the development efficiency of the study area through well layout and fracturing design.

3 Model Construction

To understand the influencing factors of $CO₂$ pre-storage fracturing and optimize the fracturing parameter design, this paper proposes a numerical simulation workflow based on the concept of geological engineering integration [[19\]](#page-19-6). The workflow consists of the following steps:

(1) Construct the reservoir geological model based seismic and well logging data;

(2) Utilize the three-dimensional refined geological model to obtain key physical parameters, such as porosity, permeability, oil saturation, Young's modulus, and Poisson's ratio. Using the geological modeling results as a foundation and constraining it with a one-dimensional geological mechanics model, achieve the accurate characterization of three-dimensional stress distribution through finite element stress simulation;

(3) Simulate the reservoir fracture propagation mechanism using an Unconventional Fracture Model (UFM) to precisely depict the pre-storage $CO₂$ fracturing network in the well area;

(4) Implement both unstructured and structured mesh techniques to represent the fracture and matrix areas in the well region. This allows for an accurate description of the fracture network morphology and proppant migration distribution;

(5) Based on the actual field engineering conditions, perform numerical simulations of key operation parameters, such as well spacing, fracturing stage length, cluster spacing, and the volume of fracturing fluid. Predict the production distribution patterns under different fracturing parameters, providing a theoretical basis for reservoir development management.

[Fig. 1](#page-3-0) is the workflow chart for Three-Dimensional Geomechanical Modeling. Through this process, we can establish a three-dimensional geological mechanics model of the work area for fracturing simulation or production history fitting and prediction. The fracture network obtained through this process of fracturing simulation is closer to the actual conditions of the reservoir site compared to the fracture network generated directly with given parameters.

Figure 1: Workflow chart for three-dimensional geomechanical modeling

3.1 Geomechanical Model Construction

3.1.1 One-Dimensional Geomechanical Modeling

The Geological Mechanics Model possesses the ability to describe the stress state and mechanical properties of rocks, which is of significant importance for conducting accurate reservoir-scale fracturing simulations [\[20\]](#page-19-7). The primary objective of constructing a one-dimensional rock mechanics model is to acquire essential mechanical parameters of rocks, including Young's modulus, Poisson's ratio, as well as the maximum and minimum horizontal principal stresses. Acoustic data serves as a crucial basis for accurately inverting reservoir mechanical property parameters. Thus, utilizing reservoir acoustic compression and shear wave velocity data and integrating methods for rock mechanics parameter inversion, it is feasible to predict dynamic Young's modulus and dynamic Poisson's ratio, which is the basis for the construction of a three-dimensional detailed geological mechanics model.

However, for wells lacking acoustic well log data, neural network methods must be employed to predict acoustic information. Simultaneously, to achieve effective conversion of dynamic and static data, corresponding conversions are necessary $[21]$. This process (shown in [Fig. 2](#page-4-0)) is primarily based on rock core acoustic experiments and triaxial stress test data. Calculations of overburden stress rely on density well log data and are computed using appropriate equations.

3.1.2 Three-Dimensional Geomechanical Modeling

Using a one-dimensional geological mechanical model as a basis, a three-dimensional mechanical model can be obtained through three-dimensional attribute interpolation. The distribution of the threedimensional geomechanical stress field in the reservoir controls the initiation and development of fractures in the reservoir. This is an important reservoir property parameter for conducting fracturing simulations. Therefore, it is necessary to carry out detailed verification of the effectiveness and accuracy of the three-dimensional geomechanical model. In this study, based on Schlumberger's VISAGE plugin, the finite numerical method is used to comprehensively consider the influence of reservoir rock heterogeneity. The three-dimensional stress field distribution in the well area is inverted and compared with field data for validation. Key parameters include the direction of the principal stress, vertical stress, horizontal stress difference, Young's modulus, and Poisson's ratio. To avoid boundary effects, the geomechanical grid is expanded laterally to three times the planar size of the geological model. Overburden, underburden, lateral layers, and stress plates are added to cover a range from the surface to a

depth of approximately 6000 m. The original grid size of the reservoir is $427 \times 206 \times 100$ (I, J, K), and after expansion, the grid size of the reservoir becomes $1443 \times 462 \times 155$ (I, J, K). In the I and J directions, the grid step sizes are 34 and 33 m, respectively, while in the K direction, it is 50 m [\[22](#page-19-9)].

Figure 2: Fracturing design process flowchart

3.2 Fracturing Simulation

In the absence of on-site operation data, a reference to the adjacent block indicates a maximum fracturing pressure of 50 MPa. A 55% slickwater ratio is used. During the pre-pad stage, a gel is employed to initiate the fractures, while during the sand-carrying stage, low sand concentration is achieved using slickwater, and high sand concentration is achieved using gel, resulting in an equivalent fluid sand ratio of approximately $9.\overline{5} \text{ m}^3/\text{m}^3$.

3.2.1 Fracture Fluid

In the hydraulic fracturing process, fracturing fluid is an indispensable working fluid. Before simulating fracture propagation, fracturing fluid characteristics such as viscosity, leak-off, friction, and gel break need to be defined. Field data indicates the use of guar gum-based gels, slickwater, and $CO₂$.

(1) Organic anti-swelling agents are added to the fracturing fluid system to enhance its anti-swelling performance and mitigate potential water sensitivity issues.

(2) The polymer slickwater system should exhibit excellent friction reduction performance to meet highvolume construction requirements. The required basic performance parameters are as shown in [Table 1.](#page-5-0)

3.2.2 Proppant

When fracturing fluids or other fluids create fractures, it is necessary to promptly transport a proppant with sufficient strength into the fractures to prevent them from closing. During on-site fracturing, 20/40 mesh silica sand is used.

	No. Components		Index				
			Low viscosity Medium fracturing fluid	viscosity fracturing fluid	High viscosity fracturing fluid	Synthetic polymer gel breaker	
1	Crosslinking performance					Cross-linked with a compatible crosslinker, forming a weak gel or a gel-like structure	
2	Gel breaking fluid	Gel Breaking Time, min	≤ 720	≤ 720	≤ 720	≤ 720	
	performance	Gel Breaking Temperature, °C	60	60	60	60	
		Apparent Viscosity of ≤ 5.0 Gel Breaking Fluid, mPa·s		≤ 5.0	≤ 5.0	≤ 5.0	
		Surface tension of Gel \leq 28.0 breaking fluid, mN/m		\leq 28.0	≤28.0	\leq 28.0	
		Interfacial tension between gel breaking fluid and crude oil, mN/m	≤ 2.0	≤ 2.0	≤ 2.0	≤ 2.0	
		Demulsification rate, %	\geq 95	≥ 95	>95	>95	
		CST ratio	<1.5	< 1.5	<1.5	<1.5	
3	Residue content (Freshwater), mg/L		$\sqrt{2}$	≤ 50	≤ 50	≤ 50	
4	Compatibility with formation water		No precipitation or flocculation observed	N ₀ precipitation or flocculation observed	N ₀ precipitation or flocculation observed	No precipitation or flocculation observed	
5		Base fluid apparent viscosity μ , mPa·s $5 \leq \mu < 20$		$20 \leq \mu < 50$	$50 \leq \mu$	$\sqrt{2}$	
6	Viscosity increase rate, %		$\sqrt{2}$	${\geq}85$	≥ 85	≥ 85	
7	Dissolution time, s		≤ 40			$\sqrt{2}$	
8		Crosslinking time, s 60° C $\leq t$ < 120 °C /				$30 - 300$	
9	Friction reduction rate, %		\geq 75	≥ 65	≥ 65	≥ 60	
10	Temperature and shear resistance	Apparent viscosity, mPa·s	$\sqrt{2}$			\geq 50 (Tested at 90°C with a shear time of ≥ 60 min and a shear rate of $170 s^{-1}$)	
11	loss, $(60^{\circ}C,$	Static leak off Leak off Coefficient, m/\sqrt{min}				$\leq 1.0 \times 10^{-3}$	
	90° C)	Initial leak off volume, / m^3/m^2				$\leq 5.0 \times 10^{-2}$	
		Leak off rate, m/min	$\sqrt{2}$			$\leq 1.5 \times 10^{-4}$	

Table 1: Polymer friction reducer performance parameters

The selection of proppant specifications is primarily based on factors such as the formation's proppantcarrying capacity, proppant conductivity, and proppant crushing rate under closure stress conditions in the target formation. Other quality parameters that the proppant used in the field must meet are listed in [Table 2](#page-6-0).

Type	Roundness Sphericity Turbidity		(FTU)	$\frac{(0)}{0}$	$\frac{10}{6}$	Acid solubility Sieve retention Crush percentage $\frac{(0)}{0}$
Badger sand ≥ 0.6		≥ 0.6	\leq 150	≤ 7.0	>90	59
Ceramic proppant	>0.7	>0.7	< 100	< 7.0	>90	<9

Table 2: Proppant performance parameter specifications for hydraulic fracturing

3.2.3 Fracture Propogation Modeling

In 2011, Kresse and colleagues [\[22](#page-19-9)] introduced an Unconventional Fracture Model (UFM) that effectively addresses the fully coupled problem of fluid flow within the fracture network and fracture elastic deformation. This model shares similar assumptions and governing equations with traditional quasi-3D fracture models: the fluid flow and the associated pressure gradient in the vertical direction are negligible except near the upper- and lower-fracture-tip region. The UFM calculates the stress intensity factors at the top and bottom tips of the fracture by considering pressure within the fracture, the geometric shape of the fracture, and in-situ stress in the formation. Subsequently, it determines the stable fracture height by matching the stress intensity factors at the tips with the fracture toughness of the layer containing the fracture tips. The specific fracture height equations [\[22\]](#page-19-9) are as follows:

$$
K_{lu} = \sqrt{\frac{\pi h}{2}} \bigg[p_{cp} - \sigma_n + \rho_f g \bigg(h_{cp} - \frac{3}{4} h \bigg) \bigg] + \sqrt{\frac{2}{\pi h}} \sum_{i=1}^{n-1} (\sigma_{i+1} - \sigma_i)
$$

$$
\times \left[\frac{h}{2} \arccos \bigg(\frac{h - 2h_i}{h} \bigg) - \sqrt{h_i (h - h_i)} \right]
$$
(1)

$$
K_{ll} = \sqrt{\frac{\pi h}{2}} \bigg[p_{cp} - \sigma_n + \rho_f g \bigg(h_{cp} - \frac{h}{4} \bigg) \bigg] + \sqrt{\frac{2}{\pi h}} \sum_{i=1}^{n-1} \left(\sigma_{i+1} - \sigma_i \right) \times \left[\frac{h}{2} \arccos \bigg(\frac{h - 2h_i}{h} \bigg) + \sqrt{h_i (h - h_i)} \right] \tag{2}
$$

$$
w(z) = \frac{4}{E'} \left[p_{cp} - \sigma_n + \rho_f g \left(h_{cp} - z \right) \right] \sqrt{z(h - z)} + \frac{4}{\pi E'} \sum_{i=1}^{n-1} (\sigma_{i+1} - \sigma_i)
$$

$$
\times \left[(h_i - z) \cosh^{-1} \frac{z \left(\frac{h - 2h_i}{h} \right) + h_i}{|z - h_i|} + \sqrt{z(h - z)} \arccos \left(\frac{h - 2h_i}{h} \right) \right]
$$
(3)

where p_{cp} represents the fracture pressure measured at the reference depth of h_{cp} , Pa; z denotes the position of the fracture, m; ρ_f is the fluid density, kg/m³; K_{Iu} and K_{II} are the stress intensity factors at the upper and lower tips of the fracture, respectively, Pa·m^{0.5}; w(z) is the fracture width at depth z, m; σ_n and σ_i represent the overburden stress at the top and in the *i*-th layer, Pa; *h* signifies the fracture height, m; h_i stands for the height from the bottom layer to the top of the *i*-th layer, m.

Furthermore, it also takes into account processes such as proppant transport between hydraulic fractures and stress shadow effects [\[23](#page-19-10)]. Compared to the Pseudo Three-Dimensional (P3D) fracture model, the Unconventional Fracture Model (UFM) considers the heterogeneity and complexity of the geological formation, enabling a more accurate representation of the geometric shape and distribution of fractures. It excels in describing the physical processes of fractures, including propagation and deformation, providing a more profound understanding of fracture behavior. Additionally, UFM allows for finer parameterization, enhancing model flexibility to adapt to different geological conditions and fracturing processes.

3.3 Reservoir Numerical Simulation

3.3.1 Grid Generation

Hydraulic fractures require grid generation prior to reservoir numerical simulation. This paper employs an adaptive unstructured grid algorithm for fracture grid generation [\[24](#page-19-11)]. Unstructured grids are used in the fracture zone and structured grids in the matrix zone, reducing the grid numbers and improving computational speed. This approach can also identify whether each fracture grid has proppant support, making the simulation results more realistic and reliable.

3.3.2 Description of Model Parameters

Permeability and porosity for the matrix grid are obtained from the geological model. The permeability and porosity of the fracture grid depend on the results of the fracturing simulation. Model parameters are presented in [Table 3](#page-7-0).

Parameter	Value	Parameter	Value
Average matrix porosity	0.16	Matrix compressibility, MPa^{-1}	1×10^{-5}
Average matrix permeability, mD	165.26	Water density, $kg/m3$	1007
Oil density, $kg/m3$	943	Water formation volume factor	1.05
Oil formation volume factor	1.03	Water viscosity, MPa·s	
Oil viscosity, MPa·s	279.4	Water compressibility, MPa^{-1}	1×10^{-4}
Oil compressibility, MPa^{-1}	1.42×10^{-3}		

Table 3: Geological model parameters

4 Model Verification

Before hydraulic fracturing design is initiated, the workflow needs to be validated. Geological mechanical parameters, such as Young's modulus, are verified through laboratory core experiments. Utilizing microseismic data and matching drawdown curves ensures the reliability of the fracturing model. Microseismic data providing information on hydraulic fractures away from the wellbore is the most widely used technique for monitoring hydraulic fractures [[25\]](#page-19-12). The reliability of geological and reservoir numerical simulation models is ensured through historical production data fitting.

4.1 The Results of the Geomechanical Model

Some one-dimensional geomechanical modeling outcomes are depicted in [Fig. 3.](#page-8-0) From left to right, these include AC (Acoustic Compression), Overburden Stress, σ_H (Horizontal Stress), Poisson's Ratio, and Young's modulus. The exact experimental values for these parameters were not directly provided in the field data. However, using empirical formulas provided on-site, these key parameters in the one-

dimensional geomechanical model can be transformed and calculated based on wellbore measurements like RT, DT, and RHOB.

Figure 3: Well-logging curves for the field (OS: overburden stress)

$$
shangfu = (2.33 + (MD - 3000)/100 \times 0.0092) \times 9.80665 \times MD/1000 \tag{4}
$$

$$
Sh = (V_{v1} + 0.033515) \times (shangfu - P_p) + P_p
$$
\n(5)

$$
AC = (DTSM - 25.06 - 30.04 \times MD^{0.15} / RT^{0.52}) / 1.13
$$
\n(6)

$$
youngMOD = DEN \times V_s^2 \times (3 \times (V_p V_s)^2 - 4) / ((V_p V_s)^2 - 1) / 1000000 \tag{7}
$$

$$
poisson = ((V_p V_s)^2 - 2)/(2 \times (V_p V_s)^2 - 2)
$$
\n(8)

where shangfu represents overburden stress, MPa; MD stands for measured depth, m; Sh signifies minimum horizontal stress, MPa; Pp represents pore pressure, MPa; AC denotes acoustic compressional, μs/m; DTSM stands for shear wave sonic log, μs/m; RT stands for resistivity, Ω·m; youngsMOD is Young's modulus, GPa;

Vs represents shear wave velocity, m/s; Vp represents compressional wave velocity, m/s; poisson stands for Poisson's ratio.

The color blocks represent the results of one-dimensional geomechanical calculations, while the lines represent the results of three-dimensional geomechanical calculations. As shown in [Fig. 4,](#page-9-0) the results of one-dimensional geomechanical calculations closely match the results of three-dimensional geomechanical calculations, providing validation for the effectiveness and validity of the geomechanical model.

Figure 4: Calibration and fitting of the key parameters

By using the Visage module to establish a three-dimensional geomechanical model [[26\]](#page-19-13), the maximum and minimum horizontal principal stresses, vertical stress, and pore pressure are corrected to match the onedimensional geomechanical model, ultimately obtaining a three-dimensional geomechanical model that aligns with the field data. As shown in [Table 4](#page-9-1), the stress state of the three-dimensional geomechanical model matches the stress state of the field's reverse fault and all data falls within the estimated range provided in the field.

Table 4: Layer moduli, poisson's ratios, and various stress mean values

Layer	Maximum horizontal principal	Minimum horizontal principal	Vertical stress,	Young's modulus,	Poisson's Pore ratio	pressure,
	stress, MPa	stress, MPa	MPa	GPa		MPa
$J1s-J1b11$	39.3	31.7	25.5	20.8	0.4	13.3
$J1b11-J1b14$	33.4	26.3	25.6	17.9	0.31	13.5
$J1b14-J1b15$	39	30.4	26.2	22.6	0.34	13.7

4.2 From Fracture Propagation to Reservoir Numerical Simulation

Structured grids make up the matrix, while fractures are represented by unstructured grids. After grid partitioning, and the parameters such as fluid, matrix type, production regime, can be set up [[27\]](#page-19-14). The production grid width is set at 5 m, the unstructured grid permeability is 3.05 mD, with dimensions of 10 m by 10 m (shown in [Fig. 5](#page-10-0)). This paper aims to demonstrate this integrated optimization method, with data sourced from field logging and neighboring well construction experiences, which may decrease the prediction accuracy.

Figure 5: Schematic representation of the grid

5 Results and Discussion

5.1 Well Spacing Design

Based on the field data, the economic limit method suggests that the horizontal section length should be greater than 500 m, while the cumulative production increment significantly decreases when the length exceeds 1200 m. Considering engineering factors such as the potential plugging after fracturing, the shallow burial depth of the studied area, and the limitations imposed by the self-weight of continuous oil pipes, it is suitable to limit the horizontal length to no more than 1200 m.

In the well spacing design, five well spacing scenarios were considered, ranging from 200 to 275 m, to explore the potential optimal well spacing range. According to the hydraulic fracturing simulation results in [Fig. 6](#page-11-0), it can be concluded that with a 260 m well spacing, the distance from the two edge wells, Well B and Well C, to the block boundary is approximately 130 m, which is half the well spacing. This configuration provides better reservoir control within the block. This observation is further confirmed in the production simulation results (shown in [Fig. 7](#page-11-1)): when the well spacing is relatively close (200, 225, 250 m), the controlled area of the three wells is relatively small, making it challenging to extract oil and gas resources at the eastern and western boundaries of the target reservoir. Conversely, when the well spacing is relatively large (275 m), due to the narrow north-south extent of the target reservoir, Well C and Well B are too close to the boundary, resulting in less effective oil and gas recovery between the wells.

Figure 6: Illustrations of hydraulic fracturing and production simulations at different well spacings (pressure contour maps)

Figure 7: Comparison of cumulative oil production at different well spacings after 10 years

5.2 Design of Hydraulic Fracturing Stage Length

Based on the field data, the economic limit suggests that the length of the horizontal section should be greater than 500 m. However, there is a significant decrease in cumulative production increment after the length exceeds 1200 m. It is necessary to consider the issue of continuous oil pipe plugging after fracturing [\[28\]](#page-19-15) and the relatively shallow burial depth of the XH018 block. The length of the horizontal section is limited by the self-weight of the continuous oil pipe. The appropriate horizontal length is suggested not to exceed 1200 m.

Taking all the factors into consideration, the appropriate length for the horizontal well in the XH018 block of the Badaowan formation is determined to be in the range of 600 to 1000 m. In the design of the hydraulic fracturing stage length, four sets of lengths are respectively set to 600, 800, 1000, and 1200 m to explore the l optimal length.

According to the hydraulic fracturing simulation results, it can be observed that as the length of the fracturing section increases, the production control range of the horizontal well also increases. Combining this with the production simulation (shown in [Fig. 8\)](#page-12-0), it can be concluded that as the length of the hydraulic fracturing section increases, the cumulative production of a single well also increases, but the rate of increase varies. The cumulative production is highest at a 1200-meter fracturing length, but the increment compared to the 1000-meter fracturing length is very small. Considering on-site construction factors, the preferred length for the hydraulic fracturing well is determined to be 1000 m. This is further confirmed by the pressure distribution map in [Fig. 9](#page-12-1), which indicates that when the fracturing length is 1200 m, there is a relatively small pressure drop around the bottom of the well, suggesting poorer reservoir properties in the western region and lower extraction efficiency.

Figure 8: Different hydraulic fracturing stage length production simulation results (using cumulative oil production as an example)

Figure 9: Pressure contour map for different hydraulic fracturing lengths after 10 years production

5.3 Cluster Spacing Design

Based on the field data, the average cluster spacing for nearby horizontal well perforations is set in the range of 20.5 to 21.6 m. Subsequently, fracturing parameter simulations were conducted, and the primary segment cluster configuration was changed from a single segment with 3 clusters to a single segment with 6 clusters. The cluster spacing was set at 10 to 12 m. Therefore, in the cluster spacing design, four sets of basic cluster spacings were considered, which were 10.52, 13.15, 17.53, and 26.3 m, corresponding to single segments with 5 clusters, 4 clusters, 3 clusters, and 2 clusters, respectively.

Fracturing simulation results (shown in [Fig. 10\)](#page-13-0) indicate that, with a consistent fracturing fluid scale, having fewer clusters in a single segment (resulting in larger inter-cluster spacing) leads to larger individual cluster injection rates and longer hydraulic fractures. Examining the pressure distribution maps reveals that when only 2 clusters are present in a single segment, the cluster spacing is too large, causing inadequate recovery near the wellbore. On the other hand, using 5 clusters in a single segment leads to excessive stimulations in the near-well zone, resulting in poor fracture communication in the far-field area, thereby impacting overall production. Similarly, this pattern aligns with the findings of Ren et al. [[29](#page-19-16)]. Taking all factors into account, the most optimal performance is achieved with a single segment containing 4 clusters.

Figure 10: Pressure cloud maps for different cluster spacing after 10 years production

5.4 Fluid Volume and Proppant Design

For the reservoir, fracturing fluid is considered an invading fluid, and any residual fluid can result in further damage to the formation, impacting production performance. Additionally, a significant amount of backflow of fracturing fluid can lead to the return of proppants, reducing the conductivity of the fractures. Therefore, in hydraulic fracturing design, it is not a matter of more is better when it comes to fracturing fluid; instead, it should be considered in combination with the actual sand body size and the backflow of fracturing fluid.

Based on field data, neighboring wells have four different pump programs with varying liquid volume capacities: single-stage programs with 450, 600, 75, and 900 $m³$ liquid volumes. Therefore, in this study, four sets of liquid volumes were designed as 450, 600, 750, and 900 m^3 to analyze their impact. In production simulation (shown in [Fig. 11](#page-14-0)), it is observed that although higher liquid volumes lead to greater cumulative oil production, there is a diminishing rate of increase in cumulative oil production as the liquid volume increases (shown in [Table 5](#page-14-1) and [Fig. 12](#page-15-0)). This is consistent with the understanding of Guo et al. [\[30](#page-20-0)].

Figure 11: Different fracturing simulation diagrams with varying liquid volumes (3D view)

Fluid	Average	Average	Average propped	Average effective	Average
volume, $m3$	fracture length	fracture height	fracture length per	propped area per	leak off per
	per cluster, m	per cluster, m	cluster, m	cluster, $m2$	cluster, m^3
450	234.95	24.98	98.22	1835.11	106.25
600	255.44	24.99	116.97	1623	150.5
750	257.23	25	128.96	1810.17	189.25
900	271.21	25.51	131.21	1915.21	227

Table 5: Statistics of various parameters of hydraulic fractures with different single-stage liquid volumes

The amount of proppant used is related to the expected fracture length, reservoir thickness, etc., and it significantly influences the geometry and conductivity of the fractures. During the fracturing process, proppants enter the fractures along with the fracturing fluid, providing support to prevent fracture closure, keeping the fractures open, and improving the reservoir's conductivity in the vicinity of the fractures. Typically, higher proppant concentrations have a noticeable effect on the fracture conductivity, leading to improved production results [\[31](#page-20-1)–[33\]](#page-20-2).

Based on field data, neighboring well have four different proppant pump programs with varying proppant volume capacities, specifically single-stage programs with 65 , 85 , 105 , and 125 m^3 proppant volumes (shown in [Table 6\)](#page-15-1). Therefore, in this study, the proppant volume was also designed as singlestage programs with 65, 85, 105, and 125 m³. Similar to the pattern observed in the liquid volume design, it is observed that higher proppant intensity leads to increased cumulative oil production.

However, beyond a certain point, increasing proppant volume results in a diminishing rate of increase in cumulative oil production (shown in [Figs. 13](#page-15-2) and [14](#page-16-0)).

Figure 12: Cumulative oil production for different liquid volumes after 10 years production

Proppant volume, $m3$	Average fracture length per cluster, m	Average fracture height per cluster, m	Average propped fracture length per cluster, m	Average effective propped area per cluster, $m2$	Average leak off per cluster, m ³
65	256.19	24.99	116.97	844.14	144
85	255.44	24.99	119.22	884.58	145.25
105	255.44	24.99	123.97	1623	150.5
125	257.7	24.99	116.22	1762.93	148.5

Table 6: Statistics various parameters of hydraulic fractures at different proppant volumes

Figure 13: Production simulation results at different proppant volumes (using cumulative oil production as an example)

Figure 14: Cumulative oil production for different proppant volumes after 10 years prodution

5.5 Design of $CO₂$ Injection Volume in the Pre-Fracturing Fluid

The relationship between the injection volume of $CO₂$ for each segment in the horizontal well is obtained based on reservoir parameters. After the injection volume for each segment exceeds 120 m^3 , the increase in cumulative production from additional injection becomes marginal. Referring to the preliminary CO_2 injection in the horizontal wells in the Ji7 area, which was around 200 m³ for alternate segments, and taking into account the on-site $CO₂$ supply and transportation capabilities, an injection volume of approximately $90-120$ m³ of $CO₂$ for each segment is determined in the horizontal well.

Therefore, this paper sets the pre-fracturing $CO₂$ injection volume for each segment as 90, 100, 110, and 120 m³, respectively. Using the 10th segment of well XHW0110 as an example for fracturing simulation, the fracturing results are shown in [Fig. 15](#page-16-1). With an increasing $CO₂$ injection volume, the fracture extension becomes longer, and the SRV (Stimulated Reservoir Volume) becomes larger. Among the four design options, the single-segment injection of 120 m³ CO₂ achieves the largest SRV, but the improvement in fracture volume from 110 to 120 m³ is relatively small (shown in [Table 7](#page-17-0) and [Fig. 16\)](#page-17-1). This is generally consistent with the relationship between pre-injection $CO₂$ liquid volume and stimulated volume studied by Tong [[34\]](#page-20-3). Considering on-site factors, it is advisable to choose 110 $m³$ of $CO₂$ injection.

Figure 15: Hydraulic fracturing simulation with different single-segment CO₂ injection volumes (3D view)

$CO2$ injection volume, m ³	90	100	110	120
Average fracture length per cluster, m	226.86	222.07	247.01	232.56
Average fracture height per cluster, m	24.98	25.54	25.48	24.95
SRV, $m3$	48225.88	49438.30	57775.46	58820.66
Average effective propped area per cluster, $m2$	613.97	886.96	1663.03	1627.67
Average propped fracture length per cluster, m	109.45	107.96	126.72	13.21

Table 7: Statistics hydraulic fracture parameters under different $CO₂$ injection volumes

Figure 16: Stimulated reservoir volume (SRV) under different single-stage $CO₂$ injection volumes

However, due to the commercial nature of the Petrel platform which lack its technical documentation regarding the mechanistic aspects of $CO₂$ as fracturing fluid, the mechanistic understanding of the effects of CO2 fracturing is not clear. This limitation introduces a certain level of constraint in our study.

6 Summary and Conclusions

This study investigates the optimization design process of pre- $CO₂$ energy storage hydraulic fracturing in heavy oil reservoir development and applies it in the Wuxia Fault Zone area on the northwestern margin of the Junggar Basin. The following are the main conclusions of this study:

- 1. The optimization design process proposed in this study, based on the Petrel platform, includes steps such as geological modeling, geomechanical modeling, hydraulic fracturing simulation, and production simulation. It contributes to the efficiency of on-site well placement and fracturing design. Through this process, reasonable fracturing parameters, such as cluster spacing, fracturing fluid volume, proppant intensity, and well spacing, can be determined to enhance production efficiency. This study utilized Petrel as an integrated simulation platform, offering enhanced visualization of simulation results compared to other platforms.
- 2. Over 60 different cases were established for optimizing well spacing, horizontal well segment fracturing length, cluster spacing, fracturing fluid volume, proppant volume, and $CO₂$ injection volume. The final optimized results are as follows: well spacing of 260 meters; horizontal well segment length of 1000 meters; cluster spacing of 13.15 meters; single-stage fracturing fluid volume of 600 m³; single-stage proppant volume of 105 m³; and single-stage CO_2 injection volume of 110 m³. These conditions resulted in the highest cumulative oil production according to production simulations.

3. This research provides a systematic optimization design method for the pre-fracturing $CO₂$ storage in heavy oil reservoirs, offering robust support for future unconventional reservoir development and CO2 management. Through the comprehensive integration of geology, geomechanics, and numerical simulation, we aim to make contributions to the sustainable development of the oil and gas industry and environmental protection.

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Availability of Data and Materials: The datasets generated during and/or analyzed during the current study are available upon request from the corresponding author. For additional materials and inquiries, please contact 202221000833@stu.swpu.edu.

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