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Optimizing tight oil extraction from low permeability tight reservoirs: a study on stress sensitivity effects and applications in carbon capture, utilization, and storage

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This paper takes into consideration the impact of stress sensitivity on bioenergy production from low permeability tight reservoirs, focusing on the post-fracturing phase. This paper established a trilinear flow model that effectively takes into account the dynamics of such reservoirs. This model incorporates stress sensitivity parameters and uses the perturbation transformation and Laplace transformation to solve the productivity prediction formula. The final productivity prediction curve post-fracturing is achieved through a numerical inversion method. This study shows that the stress sensitivity effect significantly diminishes the productivity of fractured horizontal wells used in bioenergy extraction. By introducing stress sensitivity effect parameters, this paper manages to uncover the influence law of mobility and artificial fracture parameters on productivity. Notably, as the reservoir mobility increases, the cumulative bioenergy production from fractured horizontal wells improves significantly. In addition to the above, this paper also scrutinizes the sensitivity of artificial fracture parameters and optimize both the quantity and length of these fractures. This plays a crucial role in enhancing the productivity and efficiency of bioenergy extraction from these tight reservoirs. The applicability and reliability of this method are extensively tested, thereby establishing its potential in guiding the development of low permeability reservoirs post-fracturing. Importantly, this research sets the groundwork for combining bioenergy production with Carbon Capture, Utilization, and Storage (CCUS) technologies. By focusing on optimization and stress management in tight reservoirs, this paper contributes to the sustainable production of bioenergy and reduce carbon emissions, moving a step closer to a cleaner and sustainable future.

KEYWORDS

low permeability reservoir, artificial fracture, analytic method, trilinear flow model, stress sensitivity

1 Introduction

With the advancement of unconventional oil and gas production technology, lowpermeability reservoirs have received more and more attention (Mahdaviara et al., 2022; Wang et al., 2023). In order to optimize the production of low permeability reservoirs, horizontal well fracturing in stages is the most important technical method to develop and



exploit low permeability reservoirs (Kolawole et al., 2021; Xu et al., 2021). Therefore, the production prediction of low permeability reservoirs is beneficial for designing more suitable fracturing schemes and developing low permeability reservoirs in a scientific and rational manner (Vairogs et al., 1971; Hassan et al., 2021; Wang et al., 2022).

Experimentation is a common means of studying seepage mechanisms in low-permeability reservoirs (Parekh and Sharma, 2004; Sander et al., 2017). Chen introduces a new approach to characterizing microstructure through a combination of thin sectioning, scanning electron microscopy, pressure-controlled mercury injection, rate-controlled mercury injection, and nuclear magnetic resonance (Chen et al., 2019). Kuang proposes a new NMR-based method to determine the saturation index under wet conditions of water or oil (Kuang et al., 2018). These experiments include high-pressure mercury injection, constant-rate mercury injection, nuclear magnetic resonance, as well as microscopic analysis of casting thin sections and scanning electron microscopy. Wang et al. (2018) analyze the advantages and shortcomings of these commonly used experimental techniques. Liu et al. (2020) reported the pore structure and physical properties of a low-permeability stratigraphic sample. At present, studies on the modal pore-throat structure of low-permeability reservoirs are mainly focused on dense reservoirs with controlled genesis, while studies on weakly cemented reservoirs are limited (Zhong et al.,

TABLE 1 Basic reservoir parameters.			
Parameter	Value	Parameter	Value
Reservoir size (m ³)	$800 \times 800 \times 40$	Horizontal well length (m)	800
Wellbore radius (m)	0.05	Production days (day)	300
Initial reservoir pressure (MPa)	30	Bottom hole pressure (MPa)	15
Original reservoir mobility (D/(mPa·s))	0.00002	Storage coefficient (MPa ⁻¹)	0.0004
Artificial fracture conductivity (D·cm)	15	stress sensitivity coefficient (MPa ⁻¹)	0.1
Oil compressibility factor (MPa ⁻¹)	0.0031	Oil volume factor	1
Number of artificial fractures	9	Length of artificial fracture (m)	400

2022; Li et al., 2023). Other influential work includes (Wang et al., 2021; Yongshi et al., 2021; Li et al., 2022; Zhong et al., 2022). The mechanistic study of low-permeability reservoirs has been relatively well developed, but the study of capacity prediction methods for low-permeability reservoirs still needs to be improved.

Wang developed a comprehensive data mining process to evaluate well production performance in the Monee Formation of the Western Canadian Sedimentary Basin (Wang and Chen, 2019). Due to the presence of low-permeability conglomerate, its seepage characteristics are more complex and the capacity after volume fracturing is difficult to predict. To solve this problem. Considering the trigger pressure gradient, a two-media unsteady seepage model with matrix seepage and discrete fracture network seepage was developed by (Mahmoodi et al., 2019; Vishkai and Gates, 2019) investigated sand production in dense sandstone reservoirs in terms of seepage and in situ stress distribution based on data from the Daqing oil field in the Songliao basin. Dong et al. (2019) developed a new model to study the degree of well disturbance. The model is defined by a new and original mathematical expression on production. The model proposed by Wang et al. (2020) has shown potential practical use in predicting the productivity of multistage fractured horizontal wells and in analyzing the effect of certain factors on gas production from tight, low-permeability reservoirs. The prediction method proposed by Lu and Wang (2021) can accurately predict the production of volumetrically fractured horizontal wells in the experimental area, providing some guidance for the development and adjustment of low-permeability reservoirs. To investigate the contribution of microfractures to well productivity and the effect of dynamic capillary pressure on oilwater phase flow in low permeability reservoirs (Jiang et al., 2020a; Hu et al., 2020) introduced a semi-analytical model and solved it considering the disturbance between each hydraulic fracture (Jiang et al., 2020b). Although production prediction models for lowpermeability reservoirs have been developed by previous authors from different aspects, the effect of stress sensitivity on them has rarely been considered.

This paper mainly predicts the productivity of staged fracturing horizontal wells in low permeability tight reservoirs by analytical solutions. Considering the influence of stress sensitivity on oil well production, a trilinear flow model is established. The productivity prediction formula considering stress sensitivity is solved by perturbation transformation and Laplace transformation. Finally, the productivity prediction curve after fracturing is obtained by numerical inversion





$$v = -\frac{k}{u} \left(\nabla P\right) \tag{2}$$

Among them: v the seepage velocity of the fluid in the matrix, 10^{-3} m/s; u is the fluid viscosity, mPa s; ∇P is the reservoir pressure gradient, MPa/m; k is the permeability, μm^2 ;

Formation permeability model considering stress sensitivity:

$$k = k_a e^{c_k \left(p - p_i \right)} \tag{3}$$

Where *k* represents instantaneous formation permeability; μm^2 ; k_a is the original formation permeability, μm^2 ; c_k represents the stress sensitivity factor; *p* is formation pressure, MPa; P_i is the original formation pressure, MPa.

2.2 Reservoir seepage model

The seepage of the outer reservoir is assumed to be a onedimensional single-phase flow along the x direction in the matrix, then the seepage differential equation and definite conditions of the outer reservoir are as follows:

$$\begin{cases} \nabla^2 P_o(x) = \frac{1}{\eta_o} \frac{\partial P_o(x)}{\partial t} \\ P_o|_{t=0} = P_i \\ \left(\frac{\partial P_o}{\partial x}\right)_{x=x_e} = 0, \text{ on } \Gamma_q \\ P_o|_{X=X_F} = P_I|_{X=X_F}, \text{ on } \Gamma_p \end{cases}$$
(4)

In the formula, P_o is the pressure value of the external reservoir, MPa; P_i is the initial pressure of the external reservoir, MPa; $\eta_o = \frac{k_o}{\phi \mu C_f}$ is the conductivity coefficient of external reservoir, ϕ is porosity, decimal. Γ_q represents the outer boundary and Γ_p represents the inner boundary.

The seepage of the internal reservoir is assumed to be a onedimensional single-phase flow in the matrix along the y direction perpendicular to the hydraulic fracturing. The seepage differential equations and definite conditions of the internal reservoir are as follows:

method. The sensitivity analysis of stress sensitivity coefficient, reservoir mobility and artificial fracture parameters was carried out. This study lays the foundation for integrating bioenergy production with carbon capture, utilization and storage (CCUS) technology. By focusing on optimization and pressure management in tight reservoirs, this paper contributes to sustainable production of bioenergy and reduction of carbon emissions, a step towards a clean and sustainable future.

2 Mathematical model

2.1 Basic seepage model

This paper assumes that the reservoir is horizontal equal thickness, homogeneous reservoir; reservoir top and bottom closed, only consider infinite boundary; the pressure drop caused by wellbore friction is not considered, and the pressure of each fracture is equal at the wellbore. The formation and fluid are slightly compressible, the fluid is single phase, and the flow in the formation conforms to Darcy's law.

The coupled linear flow of the trilinear flow model in three continuous flow regions is shown in Figure 1: External reservoirs beyond the tip of hydraulic fractures, internal reservoirs between hydraulic fractures, and hydraulic fractures, each region has different characteristics. Internal reservoir matrix considered homogeneous.

The state equation of reservoir is:

$$\phi = \phi_i e^{-C_{\phi}(P - P_i)}, \rho = \rho_i e^{-C_{\rho}(P - P_i)}$$
(1)

Where ϕ , ϕ_i are matrix porosity and original formation porosity, %; C_{ρ} , C_{ϕ} represent the compression coefficient of fluid and pore, MPa-1; P_i is the original formation pressure, MPa; ρ , ρ_i are the fluid density and the original fluid density, kg/m³.

The motion equation is generalized Darcy's law:







$$\frac{k_{\rm Im}}{\mu} \frac{\partial^2 P_{\rm Im}}{\partial y^2} + \frac{k_O}{x_F \mu} \frac{\partial^2 P_o}{\partial x} \Big|_{x=x_F} = C_{mI} \phi_{mo} \frac{\partial P_{\rm Im}}{\partial t}$$

$$P_I \mid_{t=0} = P_i$$

$$\left(\frac{\partial P_I}{\partial x}\right)_{y=y_e} = 0, \text{ on } \Gamma_q$$

$$P_I \mid_{y=\frac{w_F}{2}} = P_I \mid_{y=\frac{w_F}{2}}, \text{ on } \Gamma_p$$
(5)

Where, P_I is the pressure value of the internal reservoir, MPa; P_i is the initial pressure of the external reservoir, MPa; C_{mI} is the internal reservoir matrix permeability coefficient, MPa⁻¹; ϕ_{mo} is internal reservoir porosity, decimal; W_F is crack width, m.

2.3 Hydraulic fracture seepage model

The seepage assumption of hydraulic fracture is onedimensional single-phase linear seepage along the x direction in the fracture. The seepage differential equation and definite solution condition of hydraulic fracture are:

$$\frac{k_F}{\mu} \frac{\partial^2 P_{\rm Im}}{\partial y^2} + \frac{k_{tm}}{(W_F/2)\mu} \frac{\partial^2 P_o}{\partial y} \bigg|_{y=\frac{W_F}{2}} = C_F \phi_{Fo} \frac{\partial P_F}{\partial t}$$

$$P_F |_{t=0} = P_i$$

$$\left(\frac{\partial P_F}{\partial x}\right)_{x=x_F} = 0, \text{ on } \Gamma_q$$

$$P_F |_{X=0} = P_w, \text{ on } \Gamma_p$$
(6)

Where P_F is the pressure value of the internal reservoir, MPa; P_i is the initial pressure value of the crack, MPa; C_F is the fracture conductivity coefficient, MPa⁻¹; ϕ_{Fo} is the fracture porosity, decimal.

3 Numerical method

3.1 Dimensionless quantization

Define dimensionless time, distance, pressure and permeability by:



$$t_D = \frac{3.6K_I t}{(\phi c_t)_I \mu x_F^2}, x_D = \frac{x}{x_F}, P_D = \frac{P_i - P}{P_i - P_w}, C_{KD} = C_K (P_i - P_w)$$
(7)

where t_D is dimensionless time, x_D is dimensionless distance, and P_D is dimensionless pressure.

The diffusivity ratio is defined as:

$$\eta_{I} = \frac{3.6K_{I}}{(\phi c_{t})_{I}\mu}$$

$$\eta_{F} = \frac{3.6K_{F}}{(\phi c_{t})_{F}\mu}$$

$$\eta_{o} = \frac{3.6K_{o}}{(\phi c_{t})_{o}\mu}$$
(8)

In the above formula, η_I is the diffusivity of the internal reservoir, η_O is the diffusivity of the external reservoir, and η_F is the diffusivity of the external reservoir.

Dimensionless seepage differential equation for external reservoir:

$$\begin{cases} \frac{\partial^2 P_{oD}}{\partial^2 x_D^2} = e^{C_{K_{oD}, P_{oD}}} \frac{\partial P_{oD}}{\partial t_D} \\ P_{oD} \mid_{t_D = 0} = 0 \\ \left(\frac{\partial P_{oD}}{\partial x_D}\right) \mid_{x_D = x_{cD}} = 0 \\ P_{oD} \mid_{x_D = 1} = P_{ID} \mid_{x_D = 1} \end{cases}$$
(9)

The dimensionless transformation of seepage differential equations of internal reservoirs and hydraulic fractures is similar to that of external reservoirs, which is not summarized here.

3.2 Perturbation transform

Due to the strong nonlinearity of the differential equation, perturbation transformation is introduced, namely, (Khan and Wu, 2011).

$$P_D = -\frac{1}{C_{KD}} \ln\left[1 - C_{KD}\zeta_D\right] \tag{10}$$



The seepage differential equation after perturbation transformation of external reservoir is:

$$\frac{\partial^2 \zeta_{OD}}{\partial x_D^2} = \frac{1}{1 - C_{KD} \zeta_{OD}} \frac{\partial \zeta_{OD}}{\partial t_D}$$
(11)

For low permeability reservoirs, C_{KD} is generally less than 0.2, the order of magnitude of the second-order term is much smaller than the first-order solution. Therefore, the seepage differential equation after $\frac{1}{1-C_{KD}\zeta_{OD}}\approx 1$ perturbation transformation can be written as:

$$\frac{\partial^2 \zeta_{OD}}{\partial x_D^2} = \frac{\partial \zeta_{OD}}{\partial t_D} \tag{12}$$

Perturbation transformation of internal reservoir seepage differential equation can be written as:

$$\frac{\partial^2 \zeta_{\rm ImD}}{\partial y_D^2} + \frac{1}{y_{CD} C_{RD}} \left. \frac{\partial \zeta_{OD}}{\partial x_D} \right|_{x_D=1} = \frac{\partial \zeta_{\rm ImD}}{\partial t_D}$$
(13)

Perturbation transformation of fracture seepage differential equation can be written as follows:

$$\frac{\partial^2 \zeta_{FD}}{\partial x_D^2} + \frac{2}{C_{FD}} \frac{\partial \zeta_{FD}}{\partial y_D}\Big|_{y_D = W_{FD}/2} = \frac{1}{\eta_{FD}} \frac{\partial \zeta_{FD}}{\partial t_D}$$
(14)

3.3 Laplace transform and equation solving

Laplace transform is performed on the seepage differential equations of these three parts, and the following three seepage differential equations are obtained:

$$\frac{\partial^2 \bar{\zeta}_{OD}}{\partial x_D^2} - s \bar{\zeta}_{OD} = 0$$
(15)

$$\frac{\partial^2 \zeta_{\rm ImD}}{\partial y_D^2} + \frac{1}{y_{CD} C_{RD}} \left. \frac{\partial \zeta_{OD}}{\partial x_D} \right|_{x_D = 1} - s \overline{\zeta}_{\rm ImD} = 0$$
(16)

$$\frac{\partial^2 \zeta_{FD}}{\partial x_D^2} + \frac{2}{C_{FD}} \left. \frac{\partial \zeta_{FD}}{\partial y_D} \right|_{y_D = W_{FD}/2} - \frac{s}{\eta_{FD}} \bar{\zeta}_{FD} = 0 \tag{17}$$

For the seepage differential equation group (Eq. 14) of the external reservoir, after substituting its boundary conditions, the solution of Eq. 14 is:

$$\overline{\zeta}_{OD} = \overline{\zeta}_{ID}\Big|_{x_D=1} \frac{\cosh\left|\sqrt{s} \left(x_{eD} - x_D\right)\right|}{\cosh\left|\sqrt{s} \left(x_{eD} - 1\right)\right|}$$
(18)

The solution of the internal reservoir seepage equation is derived as:

$$\bar{\zeta}_{\text{Im}D} = \bar{\zeta}_{FD} \Big|_{y_D = \frac{W_{FD}}{2}} \frac{\cosh\left[\sqrt{\varepsilon_o} \left(y_{eD} - y_D\right)\right]}{\cosh\left[\sqrt{\varepsilon_o} \left(y_{eD} - \frac{W_{FD}}{2}\right)\right]}$$
(19)

Where $\varepsilon_o = \frac{\sqrt{s} \tanh[\sqrt{s} (x_{eD}-1)]}{C_{RD}y_{eD}} + s$. The solution of fracture seepage equation is:

$$\bar{\zeta}_{FD} = \frac{1}{s} \frac{\cosh\left[\sqrt{\varepsilon_F} \left(1 - x_D\right)\right]}{\cosh\left(\sqrt{\zeta_F}\right)}$$
(20)

Dimensionless transformation yield formula

$$q_D = \frac{B\mu x_F}{86.4k_F A_F (p_i - p_w)} q$$
(21)

Where B is the volume factor.

After Laplace and perturbation transformation, the solution of the yield is:

$$\bar{q}_D = \frac{1}{s - C_{KD}} \sqrt{\varepsilon_F} \tanh\left(\sqrt{\varepsilon_F}\right)$$
(22)

Where $\varepsilon_F = \frac{2\sqrt{\varepsilon_o} \tanh[\sqrt{\varepsilon_o} (y_{eD} - W_{FD}/2)]}{C_{FD}} + \frac{s}{\eta_{FD}}$. In this paper, the productivity prediction formula of trilinear

flow simulation in Laplace space has been obtained. However, to solve the production capacity in the actual space, it is necessary to calculate the solution in the real-time domain by numerical inversion. This paper adopts the numerical inversion method of AWG method (Abate and Whitt, 1992). Compared with the Stehfest algorithm, the advantages of the AWG method are Higher computational accuracy and smaller numerical error, while Stehfest algorithm produces larger truncation error in high order cases. The AWG method can handle ordinary Laplace transform, while the Stehfest algorithm can only handle bilateral Laplace





transform. The AWG method is more stable and accurate for computing the higher order derivatives of the Laplace transform. The AWG method can be easily computed in parallel, which is more conducive to engineering realization.

4 Sensitivity analysis

Firstly, the productivity curve obtained by the above derivation is compared and verified by actual reservoir. The basic parameters required for this reservoir are shown in the following Table 1. The model diagram can be seen in Figure 2. The size of the block is $800 \text{ m} \times 800 \text{ m} \times 40 \text{ m}$, and the production days are 350 days. The length of horizontal well is 800 m, the number of artificial fractures is 8, and the length of artificial fractures is 400 m. The original formation pressure is 30 MPa, and the fixed bottom hole flow pressure is 15 MPa. The stress sensitivity coefficient is considered to be 0.1 MPa⁻¹. In this paper, the productivity of fractured horizontal wells is calculated by considering the stress sensitivity coefficient and not considering the stress sensitivity coefficient, as shown in Figure 3.

As shown in Figure 3, The red circle is the real data, the blue line is the result considering the stress sensitivity, and the black line is the result without considering the stress sensitivity. Considering stress sensitivity and not considering stress sensitivity have a great influence on productivity. In this paper, the calculation of stress sensitivity is more in line with the actual production capacity, which verifies the correctness of the method. Figure 4 shows the pressure distribution at different times. It can be seen that the pressure decreases as the production proceeds. From the initial 30 MPa, it drops to about 18 MPa at the end.

4.1 Influence rule of stress sensitivity coefficient on production

From the above analysis, stress sensitivity has a great influence on productivity. Therefore, this paper selects the stress sensitivity coefficient $0-1.2 \text{ MPa}^{-1}$ (interval 0.02) for productivity prediction analysis. Other



parameters remain unchanged, as shown in the above example. The well oil production rate (WOPR) and field oil production total (FOPT) calculation results of the block are shown in Figure 5.

It can be seen from Figure 6 that as the stress sensitivity coefficient increases, the cumulative oil production of fractured horizontal wells shows a significant downward trend. When the stress sensitivity coefficient is 0, the field oil production total can reach 4.4×10^4 m³. When the stress sensitivity coefficient is 0.12 MPa⁻¹, the field oil production total is only 3.58×10^4 m³. This fully shows that the stress sensitivity coefficient has a very serious impact on productivity. Here, *HQ* is defined to represent the influence of stress sensitivity coefficient on productivity.

$$HQ = 100\% \times \frac{Q_o - Q}{Q_o} \tag{23}$$

Figure 7 below shows the influence of stress sensitivity coefficient on HQ. It can be seen from Figure 7 that HQ increases with the increase of stress sensitivity coefficient.

4.2 Effect of mobility on production

In this paper, the above examples are used. Other parameters are unchanged, and the mobility is changed to 1×10^{-6} and 1.3×10^{-5} D/ (mPa·s).

It can be seen from Figure 8 that with the increase of reservoir mobility, the propagation velocity of pressure wave increases (the supply area increases), and the cumulative production of fractured horizontal wells will increase significantly. When the mobility is less than 10^{-5} D/ (mPa·s), the production capacity increases slowly. When the mobility is greater than 10^{-5} D/(mPa·s), the production capacity increases linearly.

4.3 Influence of artificial fracture parameters

4.3.1 Number of artificial fractures

The stimulation effect of increasing artificial fracturing on fractured horizontal wells is determined by increasing contact

area, flow channel and changing seepage form, but there is also the problem of interference between fractures. Due to multi-fracture interference work cannot be realized with the trilinear model. Here the dual medium model is utilized, which is different from the trilinear flow model. In this paper, 4, 5 artificial fracturing are selected for simulation. It can be seen from Figure 9, with the increase of the number of artificial fracturing, the production of fractured horizontal wells in the early stage (before the occurrence of inter-fracture interference) is higher. With the intensification of artificial fractures will be gradually offset.

For tight oil reservoirs with stress sensitivity effect, there are more areas affected by stress sensitivity effect before the occurrence of inter-fracture interference (the time of inter-fracture interference is not affected by stress sensitivity effect), which makes more areas produce at low permeability. Therefore, with the increase of the number of artificial fractures, the cumulative production of fractured horizontal wells increases rapidly and then becomes gentle. The effect of stress sensitivity on cumulative yield is also increased.

4.3.2 Half-length of artificial fracturing

Artificial fracturing half-length is the main factor to ensure the contact area with reservoir in y direction. Because the flow of crude oil between fractures is mainly linear flow, the increase of fracture length also increases the productivity almost linearly. Then it is also affected by the action of the outer boundary. The longer the crack length, the earlier the influence of the outer boundary appears. The effect of the outer boundary depends on the mobility of the reservoir. When the mobility is low, the outer boundary has little effect on the early and middle stages of production, and the seepage process of the fluid to the fractured horizontal well system in the area exceeding the fracture length is quasi-radial flow, and its seepage resistance is much larger than the linear flow process between fractures. Therefore, increasing the length of artificial fractures, the cumulative production of fractured horizontal wells also shows a process of rapid increase first and then slow increase. In addition,

increasing the seam length can effectively weaken the negative effects of stress sensitivity (as shown in Figure 10).

4.3.3 Artificial fracture flow conductivity

Artificial fracture conductivity mainly affects the early production process of fractured horizontal wells in tight oil reservoirs, and its influence degree is directly related to reservoir flow capacity. When the reservoir flow capacity is strong, the requirement for conductivity is high; when the reservoir flow capacity is weak, the demand for conductivity is also weak. In fractured horizontal wells in tight reservoirs, the effect of conductivity on cumulative production is not as great as that of fracture number and fracture length, but the trend is the same as before. And the greater the conductivity, the higher the initial production capacity, resulting in a stronger negative effect of stress sensitivity (see Figure 11).

5 Conclusion

This paper presented a productivity prediction model for horizontal well staged fracturing in stress-sensitive tight reservoirs. The model demonstrates good adaptability in predicting the productivity of horizontal well staged fracturing. Our findings can be summarized as follows:

- This paper developed a trilinear flow model to predict the productivity of segmented horizontal wells in tight oil reservoirs, accounting for the influence of stress sensitivity. Using mathematical methods such as perturbation transformation and Laplace transformation, this paper derived the productivity prediction formula and obtained the productivity prediction curve through numerical inversion.
- 2. By comparing the productivity of horizontal wells after staged fracturing with and without considering stress sensitivity, this paper found that stress sensitivity has a significant negative impact on the productivity of fractured horizontal wells.
- 3. This paper introduced stress-sensitive effect parameters and investigated the influence of reservoir mobility and artificial fracture parameters on productivity. Results show that as reservoir mobility increases, the cumulative production of fractured horizontal wells also increases significantly.
- 4. An increase in the number of artificial fractures and the length of fractures leads to a rapid increase in cumulative production, which then gradually levels off. Although conductivity does not have as substantial an impact on cumulative production as the

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number and length of fractures, higher conductivity results in higher initial productivity, exacerbating the negative effect of stress sensitivity.

In conclusion, this study provides valuable insights into the productivity of horizontal well staged fracturing in stress-sensitive tight reservoirs. The developed model and findings can be instrumental in optimizing hydraulic fracturing operations and enhancing oil recovery in tight reservoirs.

Data availability statement

The raw data supporting the conclusion of this article will be made available by the authors, without undue reservation.

Author contributions

XJ: conceptualization, data curation, writing original draft, investigation, project administration, writing—review and editing, project administration, and supervision. GY: project administration. All authors contributed to the article and approved the submitted version.

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The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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