



Research Article

Evaluating the pore structure of low permeability glutenite reservoir by 3D digital core technology



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Abstract

In order to analyze the micro pore structure characteristics of low-permeability glutenite reservoirs more intuitively and accurately, μ -CT scanning imaging technology is used to evaluate the development characteristics of pores, fractures and high-density interstitial materials in the three-dimensional core space of samples, and the rock pore structure is analyzed from the perspective of image gray, homogeneity and pore throat connectivity. According to the maximum sphere algorithm, a digital core pore network model is extracted and established, and the probability distribution characteristics of pore radius, throat radius, pore throat ratio and core pore throat coordination number of different rock samples are studied. The results show that the reservoir space of the samples in the study area is dominated by dissolution pores and micro fractures, with pore radius ranging from 0.1 to 10 μm , throat radius from 0.1 to 7 μm , pore throat ratio from 0 to 4, and coordination number from 0 to 7. The overall pore throat ratio, throat radius and poor throat connectivity are the main reasons for the low core permeability in the study area. It is found that the 3D digital core technology can more accurately evaluate the internal heterogeneity of rocks, and the calculated pore structure parameters include both connected pore information and isolated pore information. Compared with conventional mercury injection and nuclear magnetic resonance, the 3D digital core technology has obvious advantages in the evaluation of rock micro pore structure.

Article highlights

- This paper uses the image binary segmentation method to separate the gray images of μ -CT scans of glutenite cores, and the pore space and skeleton of different rocks are characterized. And 3D gray images of 3D digital cores are constructed by PerGeos.
- The pore network model is used to analyze the reservoir pore structure, and the characteristic parameters of pore structure (such as pore radius, throat radius, pore-throat ratio, shape factor and coordination number of rock) are extracted, and the porosity and permeability are calculated to quantitatively evaluate the pore structure.
- The experiment data reveal that the main pore space of low permeability glutenite reservoir in the study area is micro-pore, with low pore coordination number, small pore-throat ratio, small pore radius and poor pore connectivity. These characteristics contribute to the low permeability reservoir, which has no obvious law of porosity and permeability, and its permeability is mainly affected by pore structure.

Keywords Digital core · Low permeability reservoir · Pore structure · Glutenite · Permian

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1 Introduction

Pore structure is an important factor affecting the property of reservoirs. Accurate acquisition of pore structure parameters and microscopic pore structure research are important components of low permeability reservoir evaluation [1]. At present, the evaluation methods of rock microscopic pore structure in laboratory are mainly 2D image observation [2] and experimental measurement [3]. The image observation method directly conducts qualitative observation and semi-quantitative statistics on the two-dimensional microscopic section of rocks by casting thin section and scanning electron microscope. This type of method is intuitive, but there are many human factors, and it is impossible to observe the distribution of pore structure in the three-dimensional space inside the rock. The experiment method mainly uses mercury intrusion and nuclear magnetic resonance experimental technology to obtain the capillary pressure curve and nuclear magnetic resonance T2 spectrum curve. The curves, which reflect the pore structure, are converted to obtain reservoir microscopic pore structure characteristic parameters. This method is mature and has high quantification degree, but it simulates the situation that external media effectively fills rock pores. Therefore, it is suitable for conventional reservoirs with good pore connectivity. Due to the large number of isolated pores in low permeability reservoirs, the experimental result has large errors. Furthermore, the experimental measurement method is not intuitive, and the core samples will be lost to some degree [4].

With continuing exploration and the complexity of geological targets, the requirements of reservoir evaluation are becoming more and more detailed. The traditional pore structure analysis methods (image observation and experimental measurement) have their own advantages and problems in describing reservoir pore types, shapes, connectivity and porosity, but qualitative analysis and quantitative evaluation cannot be carried out at the same time. X-Ray high-resolution CT imaging technology implements 3D high-speed scanning imaging of rock samples. The scanning images are used for micro-nano CT analysis, numerical reconstruction and 3D pore throat structure model of samples of different sizes. The pore throat structures in different directions are displayed on micron and nano scales, which can not only reflect the information of the whole structure but also evaluate the 3D characteristics of the inhomogeneity and micropore distribution of tiny pores. As a cutting-edge experimental method, μ -CT scanning core 3D imaging has the advantages of nondestructive core measurement, direct observation of the characteristics of connected pores and isolated pores in three-dimensional space of pore structure, and

high-precision calculation of three-dimensional space for different pore structure parameters. In recent years, it has been widely used in pore structure evaluation, especially in unconventional reservoirs such as shale oil, glutenite, low porosity and low permeability [5]. Since the 1990s, Ge Xiurun et al. used medical CT to study the fracture characteristics of rocks [6]. With the development of ray technology, Y. Jing et al. used micron CT to systematically study the fine structure of coal and rock, and obtained the spatial distribution characteristics of three-dimensional pores and fractures in coal and rock [7]. Hu Bo et al. used SEM-MAPS scanning images to establish porous media model of tight sandstone, and achieved good application results [8].

The lithology of the low permeability reservoir of Lower Permian Wuerhe Formation in Luliang Uplift of Junggar Basin is mainly glutenite with low matrix porosity. The statistical results of property analysis data show that the reservoir porosity is 5 ~ 19.1%, with an average of 8.42%, and the permeability is $(0.014 \sim 1428.2) \times 10^{-3} \mu\text{m}^2$, with an average of $2.78 \times 10^{-3} \mu\text{m}^2$. According to the analysis data of the drilled cast sheet, the pore types of the reservoir are mainly residual primary intergranular pores and secondary pores, among which intragranular corrosion pores and intergranular corrosion pores are common, with a small number of cracks. The reservoir space types are granular dissolved pores—residual primary intergranular pores and pore—crack types, with poor connectivity of interparticle seepage channels. They have the basic characteristics of low porosity and low permeability reservoirs. Previous studies on glutenite reservoirs in this area mainly focus on regional structural characteristics [9], oil and gas occurrence [10] and reservoir-forming mechanism [11]. However, there is no report on using μ -CT scanning technology to study the pore structure of this rock type. This paper introduces the technical method of micro-pore structure research of low permeability reservoir by using 3D digital core technology, and quantitatively extracts and calculates the pore structure parameters of the established visual network digital model, in order to provide technical support for the subsequent characterization of rock properties.

This paper is organized as follows. Section I introduces the principles of 3D digital core experiment. In Section II, a 3D model of the reservoir space and rock skeleton of low-permeability glutenite reservoir is established by the image binary segmentation method, and the parameters of the pore structure are quantitatively calculated and extracted by a visual network digital model. In the last section, the experiment data are analyzed in detail to determine the types, characteristics and causes of low-permeability glutenite reservoir space, so as to provide technical support for the basic research of rock properties of the target strata.

Fig. 1 Physical photos of No. 1 and No.2 rock samples from well Y bei X well



2 Methodology

2.1 Principle of CT core image processing

This experiment has been completed in key state laboratories of Southwest University of Science and Technology and the equipment used is Zeiss xradia crystal CT (Fig. 2).

It started from differences of Pore throat morphology of glutenite. A set of suitable Hole throat segmentation method was established which based on Principle of image morphology and adopting corrosion, expansion and logic algorithm.

Step 1, corrosion operator was used to eliminat throat. In image morphology, Let U be a plane area and $A1$ is a target area of U . Let $S1$ specified dimension and shape be Structural element 1. If the area represented by $S1$ located at the coordinate (x, y) is defined as $S1(x, y)$, the etching operation on the target area $A1$ can be expressed as:

$$\{(x, y) | (x, y) \in A1, S1(x, y) - A1 = \Phi\} \quad (1)$$

The size of $S1$ depends on the size of the throat in the digital core image. Through observation, it can be seen that: Since the throat can extend in any direction, the shape of the structural elements is usually disc-shaped, which can ensure that all throats can be corroded.

Step 2, use expansion operator to repair incomplete pores. In image morphology, if U is a plane area, $S2$ is a structural element 2 with a specified size and shape, and the area represented by $S2$ at coordinates (x, y) is defined as $S2(x, y)$, then the expansion operation on the target area $A2$ can be expressed as,

$$\{(x, y) | (x, y) \in u, S2(x, y) \cap A2 \neq \Phi\} \quad (2)$$

After the incomplete pores are repaired by expansion operation, the corroded boundary pixels of pores are restored; Since the expansion operation is not the inverse operation of the corrosion operation, this recovery is not accurate. In order to ensure that the pores after the expansion operation can be restored to the original pores, the diameter of $S2$ after the expansion operation is slightly larger than that of $S1$ in the corrosion operation in step 1. However,

the structure of $S2$ can be consistent with that of $S1$, and the pores after the expansion operation are $A3$.

Step 3, use intersection operation to extract original pores. The accurate pore $A4$ can be obtained by performing intersection operation between the repaired pore $A3$ and the original pore throat $A1$ through expansion operation. Intersection operation can be expressed as,

$$A4 = A1 \cap A3 \quad (3)$$

Step 4, use difference set operation to extract throat. After the accurate pore is obtained by segmentation, the pore part is removed from the original pore throat image by subtraction operation, and the accurate throat $A5$ can be obtained. Subtraction operation can be defined as,

$$A5 = A1 - A4 \quad (4)$$

2.2 3D digital core construction

The μ -CT scanning experimental data comes from well Ybei X in the work area. Samples No.1 and No.2 with pure lithology and regular shape are selected as examples to study the pore structure of low permeability glutenite reservoir. The porosity of No.1 sample is 8.8% and the permeability is $0.32 \times 10^{-3} \text{um}^2$; the porosity of No.2 sample is 7.6% and the permeability is $0.45 \times 10^{-3} \text{um}^2$. The results of routine analysis of samples are distributed within the average range of the work area, and the experimental samples are representative in the area.

In order to best identify the porosity, a cylinder with a diameter of 2 mm is drilled in the plunger sample, in which the length of No.1 sample is 28 mm and that of No.2 sample is 32 mm. The real rock sample is shown in Fig. 1. Though limited information is visible, it can be found that No.1 and No.2 rock samples are gray fluorescent gravel fine sandstone. However, there is no way to investigate the interstitial property, mineral property, pores and the distribution of pores in three-dimensional space (Fig. 2).

Two rock samples were scanned with a resolution of 1um/voxel , and 3D digital core 3D gray images were established by PerGeos digital data analysis tool. The pore space

Fig. 2 The instrument used in the experiment was Zeiss Xradia Crystal CT



development, mineral property and filling property of the core were analyzed, as shown in Fig. 3.

It can be seen that qualitative analysis of the core can be carried out by applying the rock μ -CT analysis method combined with the image gray observation image. The change of the gray value of digital core images from gray to white reflects the change of local density of the core, which is caused by different types of minerals. Therefore, the change of gray value of each image, in theory, represents a specific type of mineral. However, the density distinction of each component among different mineral types

is not obvious, so it is not easy to distinguish in the image. When the mineral types are different (such as clay and dolomitic), the density is quite different, which can cause obvious gray changes in the picture. Therefore, image qualitative analysis can determine the dominant mineral or the type of filled matter. For example, in Fig. 3, the area with higher brightness (white) is the cuttings with higher density, but it is impossible to accurately distinguish their components.

By cutting any section of the 3D digital core image, the pore structure, pore type and the distribution

Fig. 3 3D grayscale image of No. 1 and No. 2 rock samples from well Ybei X well

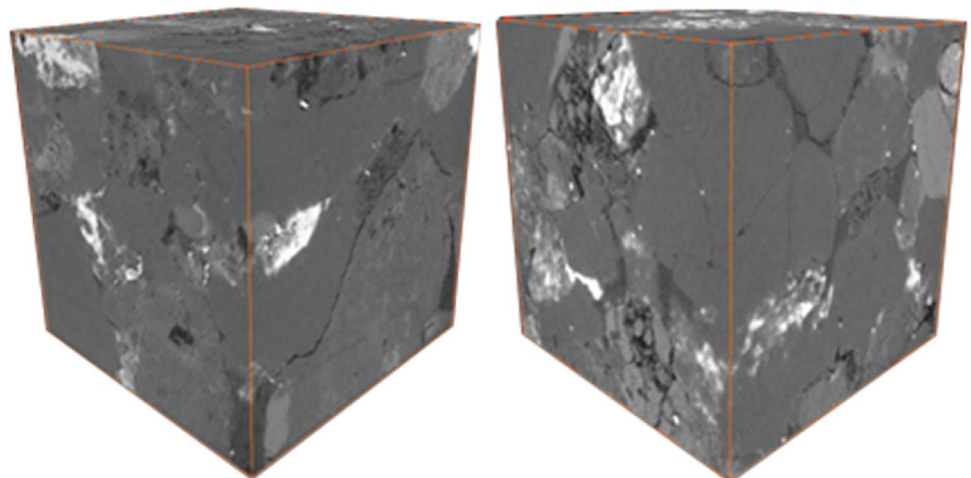


Fig. 4 2D section of 3D digital core of No.1 and No. 2 rock samples in Ybei X well

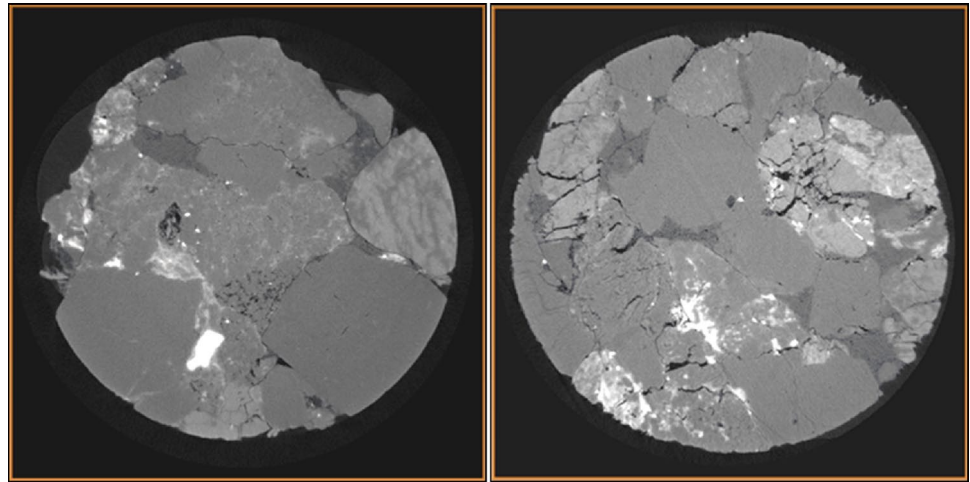
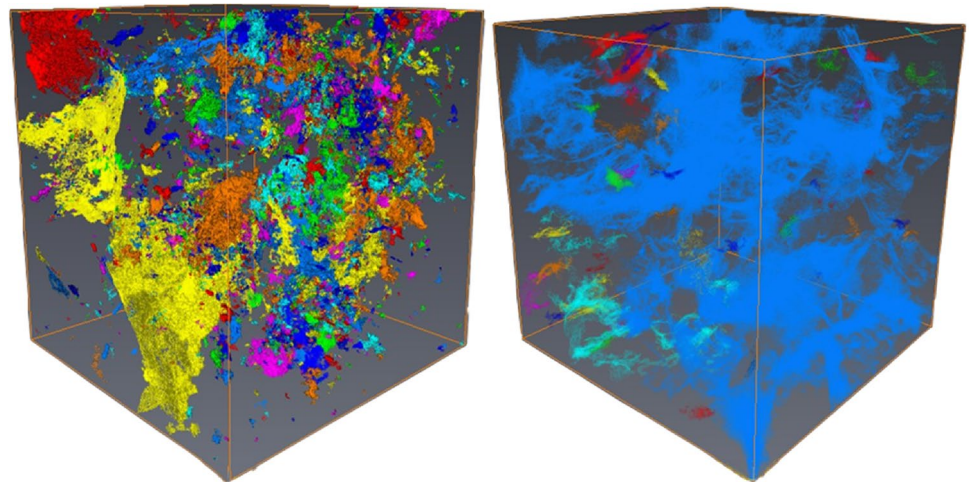


Fig. 5 Pore space of 3D digital cores of No.1 and No. 2 rock samples from Ybei X well



characteristics of fillings in 2D plane can also be obtained. The two-dimensional gray section of the rock sample can show the development degree of micro-cracks in the rock sample. Figure 4 is the two-dimensional section of the 3D digital core of No.1 and No.2 samples. From the figure, it can be seen that the cracks in the rock samples are well developed, and the gray characteristics show that the core has strong heterogeneity and there are intergranular pores at the same time. However, most of them are filled by various interstitial matters, and their micro-pores are developed. Therefore, it is concluded that the pore space of rock samples is a double pore system composed of cracks and micro-pores. From the core photo in Fig. 1, it is impossible to observe the crack development and fillings in the core. It can be seen that the core analysis by CT can detect the internal structure of the core undamaged, which provides more reliable basis for qualitative analysis.

Based on the 3D grey degree of core established by X-ray CT, through digital image processing technology,

the distribution characteristics of three-dimensional space of different components in rock are obtained. Then, the pore space of rock is identified, and the pore connectivity is analyzed. Figure 5 is a visual pore distribution diagram of rock samples in three-dimensional space. The transparent part in the diagram is the rock skeleton. The colors indicate the degree of pore connectivity. The greater the color change, the stronger the core heterogeneity and the worse the pore connectivity. It indicates that the more complex the pore structure of rock reservoirs, the stronger the heterogeneity and the poorer the connectivity. It can be seen from the figure that the pore connectivity of No.2 sample is obviously better than that of No.1 sample, and the permeability of No.2 sample is higher than that of No.1 sample from conventional analysis. In addition, 3D digital core shows that there are a large number of micro-cracks in the pore space, as shown in the yellow area of No.1 sample in Fig. 4, which leads to extremely strong heterogeneity of pores and anisotropy of rock properties.

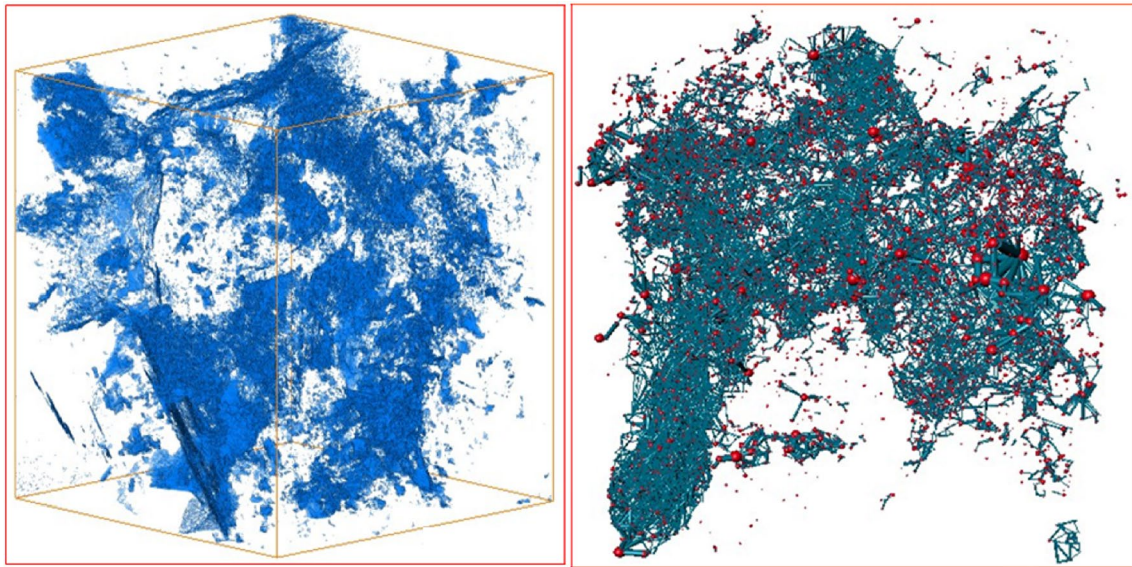


Fig. 6 Analysis chart of rock micro pore structure

3 Results and discussion

In the process of reservoir evaluation, the study of microscopic pore structure is the key of microscopic properties of reservoirs. The distribution characteristics of pores and throats and their relationship have significant influence on the reservoir performance and physical characteristics. Experiments show that 3D digital core analysis technology has obvious advantages compared with the traditional methods such as electron microscope scanning, mercury intrusion and nuclear magnetic resonance. It can make qualitative analysis and quantitative evaluation on the size and connectivity of pore space in complex reservoirs.

In this paper, the pore network model is used to analyze the pore structure. The basic idea is to extract structured pore and throat models from the 3D digital core image after threshold segmentation by mathematical algorithm to represent the pore space. The “maximum sphere” algorithm is the main method to extract 3D digital core pore network model at present [12]. It is assumed that a sphere is placed in each pore (including throat), and it has a largest possible diameter, so the pore space will be represented by the sphere with the largest size. The throat is a connected structure with a certain length, and the throat structure is characterized by interconnected balls with a smaller diameter. Similar to tubular structures, the number of tubular structures connected by each “largest ball” is the coordination number. Finally, the pore space of 3D digital core can be simplified to a “spherical tube model” with pore and throat as units.

Based on the pore space shown in 3D core (Fig. 6) and rock microscopic pore structure analysis established by

X-ray CT, the pore network model is extracted by using the maximum sphere algorithm. In the pore network on the right side of Fig. 5, the red is pore body and the green is throat. Based on the established digital core model, the pore space in the sample can be clearly extracted, the parameters such as porosity and permeability of the sample can be calculated, and the characteristic parameters (pore radius, throat radius, pore throat ratio, and shape factor) can be counted.

Porosity is the percentage of pore space in the total volume of the sample, which is expressed as

$$\phi = \frac{V_p}{V_b} \times 100\% \tag{5}$$

where ϕ is porosity (%), v_p is the pore volume, and v_b is the total volume.

In the 3D digital core model, for the continuous part, the pore volume is

$$V_p = \int I(x, y, z) dx dy dz \tag{6}$$

For the discrete part, the pore volume is

$$V_p = \sum_{i,j,k} I(x_i, y_i, z_i) \tag{7}$$

The total volume of the sample is

$$V_b = g_x \times g_y \times g_z \tag{8}$$

Combining the above equations, the porosity of digital core samples can be obtained.

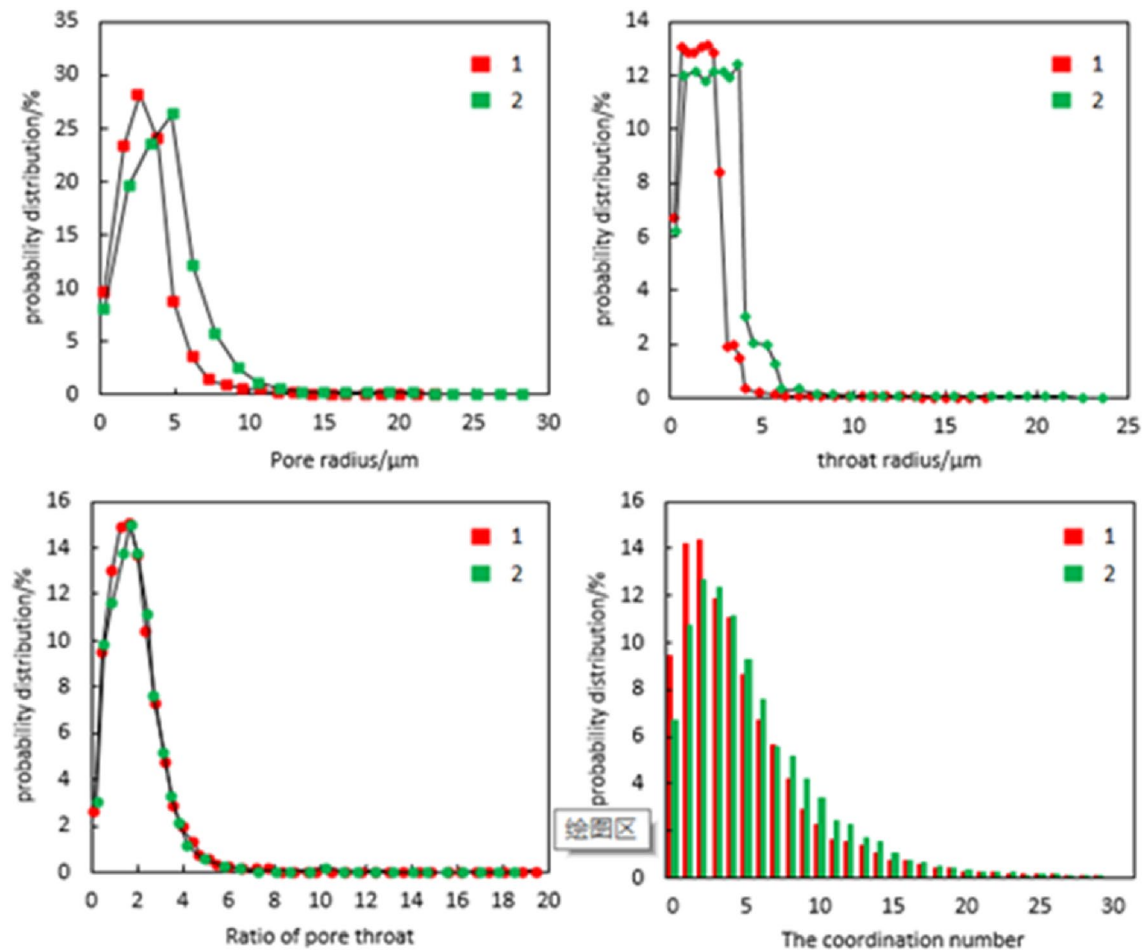


Fig. 7 Probability distribution of pore structure parameters of No.1 and No.2 rock samples from Ybei X well

Pore radius is an important parameter to measure the size of a single pore. When the pore is a regular circle, the radius of the “maximum sphere” can be regarded as the pore radius, and at this point, a sphere can fill the pore. However, the spatial shape and structure of the pores are complex. When the pores are irregular, two or more overlapping “maximum sphere” are needed to make the pores filled to the maximum extent. In this case, the pore radius should take the middle value of the radii of two adjacent balls.

Throat radius is an important parameter to calculate the critical seepage pressure of fluid flowing through the throat, and it is the basis to evaluate the pore structure of reservoirs as well. The throat radius is also calculated by the “maximum sphere” filling the throat space, which is the radius of the largest sphere placed there.

Pore structure characteristic parameters such as pore-to-throat ratio, shape factor and coordination number can be extracted according to the distribution of pores and channels that have been determined in the pore network model.

The evaluation of rock pore structure can be divided into qualitative and quantitative aspects. The qualitative evaluation indexes include the development characteristics of pores, cracks and high-density fillings, which are obtained from 3D gray images and 2D sections scanned by μ -CT. Quantitative evaluation is to obtain the characteristic parameters of the rock pore structure. The experimental parameters of the rock pore structure are used to analyze the microscopic pore structure of reservoirs. Figure 7 shows the probability distribution of pore radius, throat radius, pore-to-throat ratio and the pore throat coordination number for No.1 and No.2 samples.

It can be seen from the figure that the pore radius of No.1 sample is 0.1 ~ 7.1 μm , and the pore radius of 2.4 μm accounts for the largest percentage, reaching 28%. The throat radius is 0.1 ~ 4.1 μm , and the throat with a radius of 1.1 ~ 3.0 μm accounts for the largest percentage, around 13%. The pore-to-throat ratio of this sample is 0.11 ~ 4.0, and most of the samples have a ratio of 2. The coordination number (the number of throats connected with a pore) is predominantly between 0 and 15, and the cumulative proportion is about

37.5% when it is 0 to 3. The pore radius of No.2 sample is 0.1 ~ 10 μm , and the pore radius of 5.5 μm accounts for the most, reaching 26%. The throat radius is 0.1 ~ 6 μm , and the throat with a radius of 1.1 ~ 5.1 μm accounts for the most, about 12%. The pore-to-throat ratio of this sample is 0.1 ~ 4.0, and the ratio of most samples is 2.5. The coordination number is predominantly between 0 and 15, and coordination number of 1 to 6 accounts for about 56%.

The results show that the pore structure parameters and the distribution patterns of the two independently extracted samples have little difference, which indicates that the experimental results are not random. The pore radius and throat radius of No.2 sample are larger than those of No.1 sample. The pore-to-throat ratio of No.2 sample is slightly larger than that of No.1 sample, and the coordination number of No.2 sample is obviously higher than that of No.1 sample. This reveals that the pore structure of No.2 sample is better than that of No.1 sample, and the pore-throat connectivity of No.2 sample is slightly better. However, the average pore radius of sample 1 is 1.8 μm , and that of sample 2 is 2.2 μm . There is little difference in pore structure between sample 1 and sample 2. The results fully show that the error of conventional pore structure analysis is large, while the accuracy of core digital scanning is high. Comprehensive comparison demonstrates that the tight reservoir has poor pore connectivity and low coordination number. Because the main pore space of rock is microfractures, the pore-to-throat ratio is small and the pore radius is small. These factors are the main reasons for the low permeability of the reservoir. According to the conventional property analysis, the porosity of No.2 sample is 7.6%, the permeability is $0.45 \times 10^{-3} \text{um}^2$, and the permeability of the sample is slightly higher than that of No.1 sample of $0.32 \times 10^{-3} \text{um}^2$. The results of CT scanning pore structure analysis are in good agreement with the conventional experiment, which shows that the porosity and permeability of tight reservoirs are inconsistent and the permeability is mainly controlled by the pore structure. According to the conventional property analysis, the porosity of No.2 sample is 7.6%, and the permeability is $0.45 \times 10^{-3} \text{um}^2$, slightly higher than that of No.1 sample ($0.32 \times 10^{-3} \text{um}^2$). It indicates that the porosity and permeability of the low permeability glutenite reservoir are inconsistent, and the increase of porosity results in decreased permeability. The analysis of digital core experiment results demonstrates that the permeability of low permeability glutenite reservoir is greatly influenced by pore structure.

4 Conclusion

The low permeability glutenite reservoir has various pore types, and it is difficult to characterize the pore structure. The 3D digital core technology can obtain the

three-dimensional distribution and connectivity of pore throats without damaging the internal structure. Besides, it can accurately locate the connected pore throats and isolated pore throats of different size samples, and analyze the pore structure parameters quantitatively. In this paper, the 3D digital core technology is employed to analyze the pore structure of low permeability reservoir qualitatively and quantitatively. It is found that the reservoir space is a dual pore system composed of fractures and micro pores. The main reasons of "low permeability" of the reservoir are the low coordination number, small pore-throat ratio, small pore radius, and poor connectivity. The scanning accuracy of micron-scale digital core technology is high, which is beneficial to the evaluation of microscopic pore structure. However, this technique has high cost and small visual threshold area, so it is necessary to select very representative samples for analysis.

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Declarations

Competing interests The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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