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Pore structure and its impact on susceptibility to coal spontaneous combustion based on multiscale and multifractal analysis

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The relationship between the properties of coal and its tendency to spontaneous combustion is critical for the environment, safety concerns, and economy. In this study, to eliminate the complex influence of moisture; the samples having similar moisture content were selected from Shanxi and Henan provinces. The chemical properties, physical properties, and tendency of coal samples to spontaneous combustion were characterized based on the conventional analysis, mercury intrusion porosimetry, fractal dimensions, and crossing point temperature (CPT). The results confirmed that the coal rank, volatile matter, oxygen contents, and fixed carbon content had a good linear relationship with the CPT. The relationship between the ash content and CPT presented a “U-shaped” non-linear correlation. For the pore size distribution, the total pore volume also possessed a linear positive correlation with the CPT. The fractal curves could be distinctly divided into two stages: low-pressure (<20 MPa) and high-pressure (>20 MPa), from which the fractal dimensions were obtained using the Sponge and Sierpinski models. The relationship between the fractal dimensions (D_{s1} , D_{s2} , and D_{g1}) and CPT could be divided into two distinct stages: a decrease in the CPT with increasing fractal dimensions (2.6–2.85), and then an increase in the CPT. CPT decreased with increasing parameters of D_1 , D_2 , H , and D_{10} , and it gradually increased with increasing D_{-10} – D_{10} , D_{-10} – D_0 , and D_0 – D_{10} . The above characteristics are important to comprehensively and systematically reveal the mechanism of spontaneous combustion.

Low temperature (i.e., <200 °C) chemical and/or physical processes in fossil fuels result in the accumulation of heat, which thereby leads to spontaneous combustion^{1,2}. Numerous problems are caused by spontaneous combustion of coal, such as environment issues (emission of greenhouse gases, toxic and harmful gases, and trace elements)^{3–6}, safety concerns (dust and gas explosions, unstable overlying rocks, large cracks, and subsidence)^{7–9}, and economical loss (destruction of apparatus and equipment, and loss of coal resources)^{3,10}. Therefore, it is important to evaluate the degree of proneness for the prevention of the spontaneous combustion of coal.

The crossing point temperature (CPT) is an index to evaluate the tendency of coal to spontaneous combustion^{11–13}. Coal properties, oxygen, and ignition are three main factors for predicting the spontaneous combustion tendencies of different coals in chemical and/or physical processes¹⁴. Numerous studies have been reported on the intrinsic properties of coal and its spontaneous combustion tendency^{15–20}. Qi *et al.*²¹ analyzed the relationship between the CPT and moisture, coal rank, and sulfur content. Nimaje and Tripathy²² demonstrated that the parameters of the ultimate analysis had a significant correlation with the Olpinski index, and this relationship could be used as a reliable index to assess the susceptibility of Indian coals to spontaneous combustion. Chandra and Prasad²³ found that low coal horizons (less than 36% V.M.; Romax. 0.90–1.25%) were least susceptible to spontaneous combustion. Pattanaik *et al.*²⁴ demonstrated that the intrinsic properties of coal (stratigraphy or coal rank, volatile matter, and petrography) had a good relationship with the susceptibility indices of spontaneous combustion (CPT, and differential thermal analysis/DTA). Nimaje *et al.*¹³ analyzed the statistical relationship

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between the CPT and the proximate analysis parameters, and found that mixture surface regression (MSR) model was more effective in predicting spontaneous heating liability of coal.

The influence of water on the self-heat process is complex²⁵. Moisture can accelerate the oxidation, and enhance the thermal response by hindering the formation of stabilized radicals²⁵. However, the liability of spontaneous combustion of air-dried coal samples increases in the moisture content¹⁸. Zhang *et al.*²⁶ and Qi *et al.*²¹ found that moisture may play a role in slowing down spontaneous combustion.

Several pore classification standards have been proposed, which are as follows: 1) Ходот²⁷ classified pores into micropores (<10 nm in diameter), transitional-pores (10–100 nm), mesopores (100–1000 nm), and macropores (>1000 nm); the International Union of Pure and Applied Chemistry (IUPAC) has classified pores into micropores (<2 nm), mesopores (2–50 nm) and macropores (>50 nm); and Yao *et al.*²⁸ classified pores into adsorption-pores (<100 nm) and seepage-pores (100 nm).

Pore structures play an important role in the combustion and oxidation of coal^{1,29–31}. Pores provide a good pathway to the transport oxygen to coal surface at low temperature, which increases consumption of oxygen^{32,33}. Small space or volume can improve the material performance³⁴. Karsner and Perlmutter³⁵ found that coal with a large pore volume had high oxidation rate. Parsa *et al.*¹ evaluated the effect of densification on brown coal on its physical properties and spontaneous combustion propensity. The results exhibited that the decrease in micropore volume led to a decrease in the contact between oxygen and the coal surface, thereby the increasing the critical ignition temperature. Air can more easily contact the macropore structures of coal (the decreased in the reactivity is initially rapid) than the micropore structures³⁰. Mainly investigations on coal structural characteristics such as the pore volume, surface area, pore size distribution, and heterogeneity are conducted to reveal the spontaneous mechanism.

Pore structure characteristics have been widely applied in Coalbed Methane (CBM) exploration and exploitation^{36–39}, shale gas reservoirs, and predicting favorable zones^{40–42}. Mono fractal approaches have been extensively used in investigating the fractal characteristics of pore structures^{28,43,44}. However, a single-scale (mono-fractal) analysis or a single fractal dimension cannot explain the differences in the pore size intervals and types of erratic variation or local variation that occur in the inner distribution of pore sizes⁴⁵. Therefore, a multi-scale fractal model is used for different coal ranks, and tectonically deformed coals with the same coal rank^{37,46}. Besides, multifractal methods are also used for non-homogeneous porous media to reveal the heterogeneous pore characteristic^{47–51}. Li *et al.*⁵¹ and Song *et al.*⁴⁵ conducted a the multifractal analysis to investigate the variability and heterogeneity of tectonically deformed coals by high-pressure mercury intrusion experiment. Therefore, multifractal analysis is an efficient method to investigate pore size distribution. Nevertheless, focus on the relationship between the spontaneous combustion characteristics and heterogeneity in different coal ranks is lacking.

The objective of this study was to develop an understanding of the relationship between the properties of coal and its tendency to spontaneous combustion. The experiments were conducted by conventional analysis, mercury intrusion porosimetry, and CPT. To eliminate the complex influence of moisture in the process of spontaneous combustion, coal samples having similar moisture contents were selected. This was done to reveal the coal intrinsic properties (coal rank, volatile matter, oxygen contents, fixed carbon contents, and sulfur content), pore characteristics (pore volume, surface area, pore size distribution, and heterogeneity), and tendency of coal to spontaneous combustion. Both multiscale and multifractal methods were used to systematically investigate the heterogeneity and susceptibility to coal spontaneous combustion. Such research is important to comprehensively and systematically reveal the mechanism of spontaneous combustion.

Experiments and Modelling

Geological setting. A total of seven coal samples were collected from Shanxi and Henan provinces (Fig. 1), which contain the major coalfields in China. Six samples were collected from three coalfields in Shanxi, including Ningwu, Xishan, and Qinshui, and one sample was collected from Henan coalfield (Pingdingshan). The main coal-bearing stratum of these coalfields belongs to the Carboniferous–Permian system. Samples were directly collected from coal colliery following the Chinese Standard Method GB/T 19222–2003 and were carefully packed and taken to the laboratory for experiments.

Samples and experiments. Seven coals were collected from Ningwu, Dongqu, Guandi, Longle, Fenghuangshan, and Pingdingshan collieries in Shanxi and Henan provinces. The coal samples were crushed to –200 meshes (~0.074 mm) and vacuum-dried. The ultimate analysis and proximate analysis were conducted in accordance with the Chinese national testing standards (GB/T 212–2008 and GB/T 31391–2015). Vitrinite random reflectance (%Rr) were measured for all the coal samples on the same polished sections using a Leitz MPV-3 photometer microscope, following conventional methods in accordance with the Chinese national standard (GB/T 6948–1998).

The CPT measurements were conducted at the China University of Mining and Technology. The details of the experiment are reported in our previous work²⁶. The samples (mass 50 g with a particle size of ~0.18–0.43 mm) were exposed to a dry air flow of 50 mL/min within the reactor with a temperature ramp rate of 1 °C/min with coal and oven temperatures being recorded. When the coal temperature is equal to the oven temperature, the sample is at the CPT^{21,52}.

The coal samples were prepared by a vacuum drying for 12 h at 70–80 °C. High pressure mercury injection (HPMI) experiment was performed for the samples using an Auto Pore IV 9510 HPMI instrument at the China University of Mining and Technology. The mercury injection pressure ranged from 0.90 to 4.0×10^4 PSIA. The pore diameters obtained were from 5.35 to 2.28×10^5 nm.

Fractal dimensions. *Sierpinski model.* Based on the Sierpinski model⁵³, the fractal dimensions (D_s) can be calculated by the following equation:

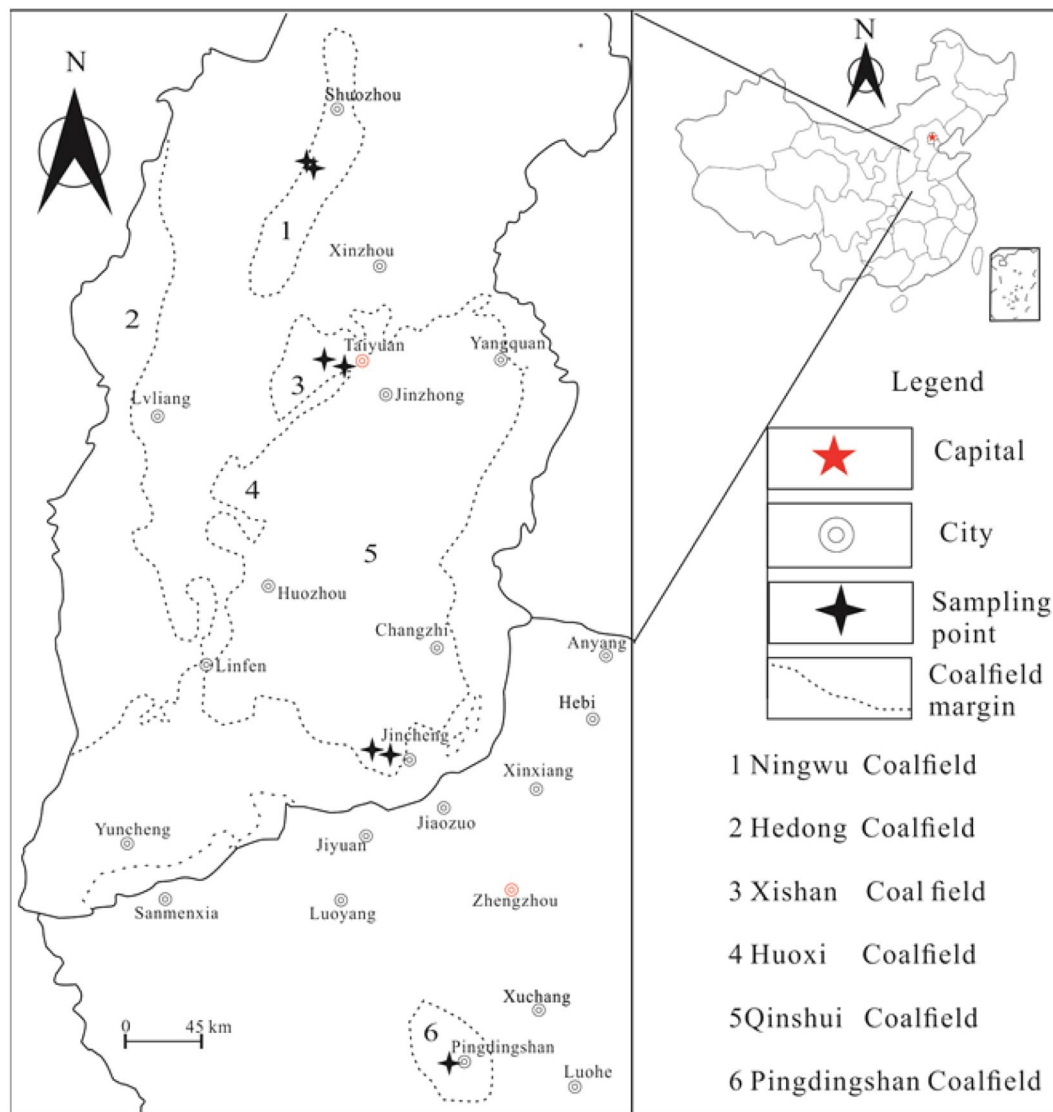


Figure 1. Positions of the research areas and the distribution of the sampling points.

$$\ln(V) = (3 - D_s)\ln(P - P_t) + \ln\alpha \tag{1}$$

where V is the mercury injection amount at P in mL, D is the volume fractal dimension, P_t is the mercury inlet pressure in MPa, and α is a constant. If the slope of the $\ln(P - P_t)$ vs. $\ln(V)$ curve is K , then D_s is $3 - K$.

Sponge model. Based on the Menegr model⁵³, the fractal dimensions (D_g) can be calculated by the following equation:

$$D_g = \ln[dVP(r)/dP(r)] - \ln\alpha/\ln P(r) + 4, \tag{2}$$

where $V_{P(r)}$ is the cumulative injection volume at a given pressure $P(r)$ and α is a constant. Therefore, the pore fractal dimension, D_g , can be obtained by: $D_g = 4 + A$, where A is the slope in Eq. (2).

Multifractals. Multifractal analysis is used to measure the statistic Hg pore size distributions. The pore diameter interval (l) ranging from 0.006 nm to initial diameter (responding to the least pressure) was selected to generate a box ($N(\epsilon) = 2^k, \epsilon = L \times 2^k$) by dyadic partitions in k stages ($k = 1, 2, 3, \dots$), where L is the length of the support^{45,48,50,51}. The probability, $P_j(\epsilon)$, of the pore size can be calculated as⁴⁸,

$$P_j(\epsilon) = \frac{N_j(\epsilon)}{N_t}, \tag{3}$$

where, $N_j(\epsilon)$ is the volume of a box ($j=1, 2, 3, \dots$), and N_t is the total volume of the system. The probability for each box of size ϵ unit can be calculated as⁴⁸,

$$P(\varepsilon) = \varepsilon^{-\alpha_j}, \quad (4)$$

where α_j is the coarse Hölder or singularity exponent for the boxes, which theoretically represents how the singularities of a system tend to infinity in the limit $\varepsilon \rightarrow 0$.

The α exponent, N_α , was used to evaluate the number of boxes, as follows:

$$N\alpha(\varepsilon)\alpha^{-f(\alpha)}, \quad (5)$$

where the set of $f(\alpha)$ values represents the spectrum of fractal dimensions that characterize the abundance of the set of points with singularity α . $f(\alpha)$ can be calculated as⁴⁸:

$$a(q) \propto \frac{\sum_{j=1}^{N(\varepsilon)} u_j(q, \varepsilon) \log\{p_j(\varepsilon)\}}{\log(\varepsilon)} \quad (6)$$

and

$$f\{\alpha(q)\} \propto \frac{\sum_{j=1}^{N(\varepsilon)} u_j(q, \varepsilon) \log\{u_j(q, \varepsilon)\}}{\log(\varepsilon)}, \quad (7)$$

where $\mu_j(q, \varepsilon)$ and $P_j(\varepsilon)$ are the normalized measures, defined as⁴⁸:

$$\mu_j(q, \varepsilon) = \frac{p_j^q(\varepsilon)}{\sum_{j=1}^{N(\varepsilon)} p_j^q(\varepsilon)} \quad (8)$$

$$\chi(q, \varepsilon) = \sum_{j=1}^{N(\varepsilon)} p_j^q(\varepsilon), \quad (9)$$

where $\chi(q, \varepsilon)$ can be calculated as:

$$\chi(q, \varepsilon) \propto \varepsilon^{-\tau(q)}, \quad (10)$$

where $\tau(q)$, q th mass exponent, can be defined as⁴⁸,

$$\tau(q) = \lim_{\varepsilon \rightarrow 0} \frac{\log \mu(q, \varepsilon)}{\log\left(\frac{1}{\varepsilon}\right)} = (1 - q)D_q, \quad (11)$$

where D_q called the generalized fractal dimensions or Rényi dimensions can be calculated as⁴⁸,

$$D_q = \frac{1}{q-1} \lim_{\varepsilon \rightarrow 0} \frac{\log \mu(q, \varepsilon)}{\log \varepsilon} (q \neq 1), \quad (12)$$

$$D_1 = \frac{\sum_{j=1}^{N(\varepsilon)} p_j \log p_j}{\log(\varepsilon)} (q = 1), \quad (13)$$

Results and data analyses

Conventional characteristics. The vitrinite random reflectance (Rr, %) of the coal samples was ranged from 0.58% to 3.43%, corresponding to medium-rank coal (bituminous coals A, B, C, and D) to high-rank coal (anthracite coals B and C) (ISO 11760, 2005)⁵⁴. Proximate analysis showed that the volatility of the coals varied from 5.95% to 43.45%. The moisture content was similar in different coal ranks. The fixed carbon ranged from 27.83% to 80.88%. The total sulfur content changed from 0.42% to 9.36%. The ash yields also had a wide range, varying from 5.47% to 49.09% (Table 1).

Macropores structure from HPMI experiments. *Pore structure distribution.* Table 2 lists the pore parameters obtained from the HPMI. In this study, the classification standards defined by Yao *et al.*²⁸ and the sharpness of the curve were used. The following three ranges are present: $V_1 < 100$ nm, $100 < V_2 < 1000$ nm, and $V_3 > 10000$ nm. Samples S_2 and S_1 have the largest and smallest cumulative pore volume in V_1 , respectively. Cumulative pore volumes V_2 have a relatively lower distribution than V_1 or V_3 , and among the samples, sample S_3 has the highest distribution. The coal rank ranges from bituminous A to anthracite B when the cumulative pore volume, V_2 , is the same (0.002 cm³/g). Samples S_1 and S_3 have the smallest and largest cumulative pore volumes V_3 , respectively, which are in response to the total pore volume.

The mercury injection and withdraw curves as well as the pore size distribution are shown in Fig. 1a,b, respectively. For the high-rank coals, the mercury intrusion and extrusion curves (Fig. 2a) display a similar trend (parallel type), indicating the dominance of the parallel plate pores and a good connectivity for gas diffusion. The shape increases in case of mercury intrusion at low pressures reveal a high proportion of V_3 , and the slightly straight lines reveal a poorly developed V_1 , which are in agreement with pore volume distribution. For the medium-rank coals,

Samples	Coal coalfield	Rr (%)	Coal ranks	Proximate analysis, %				Ultimate analysis, %				
				M _d	A _d	V _d	FC _d	S _{t,d}	O _{daf}	C _{daf}	H _{daf}	N _{daf}
S ₁	Ningwu	0.58	Bituminous D	1.02	49.09	23.07	27.83	9.36	12.36	63.23	4.95	1.08
S ₂		0.63	Bituminous C	1.01	5.47	43.45	51.09	2.51	8.49	81.83	5.66	1.37
S ₃	Pingdingshan	1.31	Bituminous B	1.00	38.81	18.17	43.02	0.42	3.69	80.61	4.43	1.57
S ₄	Dongqu	1.59	Bituminous A	1.01	6.47	14.82	78.79	1.78	4.76	87.88	4.13	1.32
S ₅	Guandi	2.13	Anthracite C	1.01	9.40	9.71	80.88	1.79	3.36	89.73	3.69	1.23
S ₆	Longle	3.08	Anthracite B	1.02	26.61	7.04	66.35	4.94	3.89	86.13	2.52	0.74
S ₇	Fenghuangshan	3.43	Anthracite B	1.01	16.03	5.95	78.02	5.47	0.85	89.28	2.61	0.75

Table 1. Properties of the coal samples having different coal ranks.

Samples	Cumulative pore volume (cm ³ /g)			Mean volume (cm ³ /g)	Mean surface area (m ² /g)	Total pore volume (cm ³ /g)
	V ₁	V ₂	V ₃			
S ₁	0.011	0.009	0.006	0.018	0.306	0.026
S ₂	0.039	0.003	0.024	0.030	0.369	0.066
S ₃	0.015	0.014	0.005	*	*	0.033
S ₄	0.024	0.002	0.105	0.115	0.158	0.130
S ₅	0.021	0.002	0.115	0.120	0.062	0.137
S ₆	0.015	0.002	0.075	0.082	0.097	0.092
S ₇	0.019	0.002	0.202	0.020	0.204	0.223

Table 2. Results of the HPMI and pore distribution for different coal ranks. Note: V₁, V₂, and V₃ are the cumulative pore volumes of <100 nm, 100–10000 nm, and >10000 nm respectively, and V_t = V₁ + V₂ + V₃. *no data.

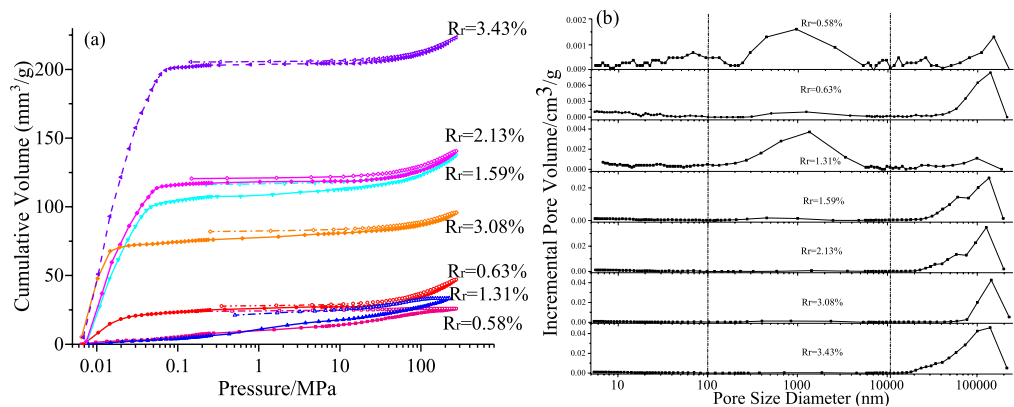


Figure 2. (a) Mercury injection and withdraw curves, and (b) Pore size distribution.

there is larger space between the mercury intrusion and extrusion curves (hysteresis loop) (Fig. 2a) than between those of the high-rank coals (tip-edge type), indicating a larger V₁, small V₂ and V₃, and poorer pore connectivity than that in the high-rank coals. The larger hysteresis loop suggests a significant existence of the semi-closed pores. The pore size distribution curves (Fig. 2b) remarkably change with the increase in the coal rank.

Fractal dimensions by Sierpinski model. The fractal curves based on the Sierpinski model are presented in Fig. 3. Fractal dimension D_{s1} and D_{s2} are obtained in low-pressure (<20 MPa, responding to the see page pores, >100 nm) and high-pressure (<20 MPa, responding to adsorption pores, <100 nm) stages. The correlation index of D_{s1} ranges from 0.30 to 0.99. Medium-rank coals S₁, S₂ and S₃ and high-rank coal S₅ present a better correlation. However, the correlation index of D_{s2} had a high correlation index (0.80–0.995). The values of D_{s1} and D_{s2} ranges from 2 to 3, indicating their power law relationship with the fractal pore surface. The values of D_{s1} and D_{s2} are 2.64–2.98 (2.90 in average) and 2.80–2.98 (2.88 in average), respectively.

Fractal dimensions by sponge model. The fractal curves are divided into two stages based on the classic geometry model (sponge model) (Fig. 4), and thus two fractal dimensions (D_{g1} , low-pressure and D_{g2} , high-pressure) are obtained using the mercury intrusion data. D_{g1} exhibits a good linear relationship (correlation index R^2 ,

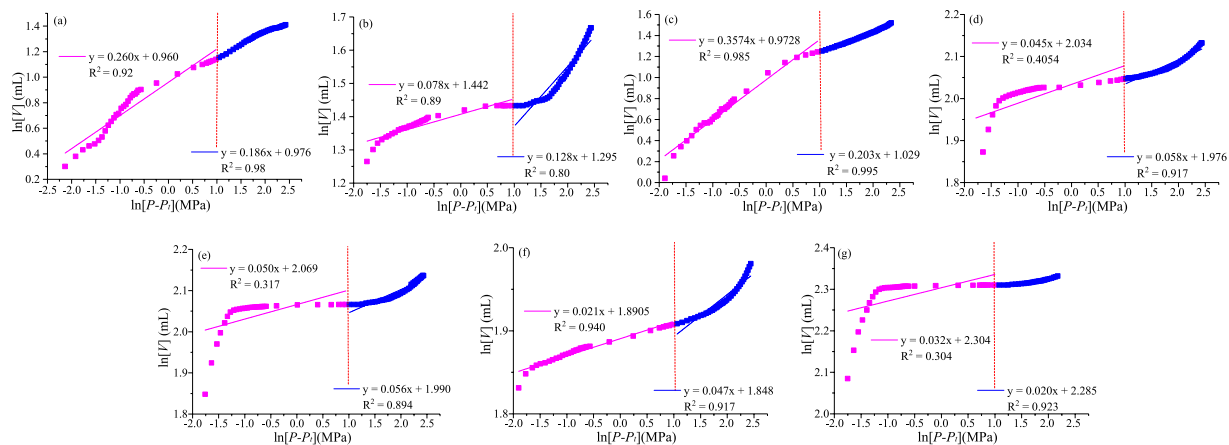


Figure 3. Fractal dimensions D_{s2} and D_{s1} for different coal ranks based on the Sierpinski model.

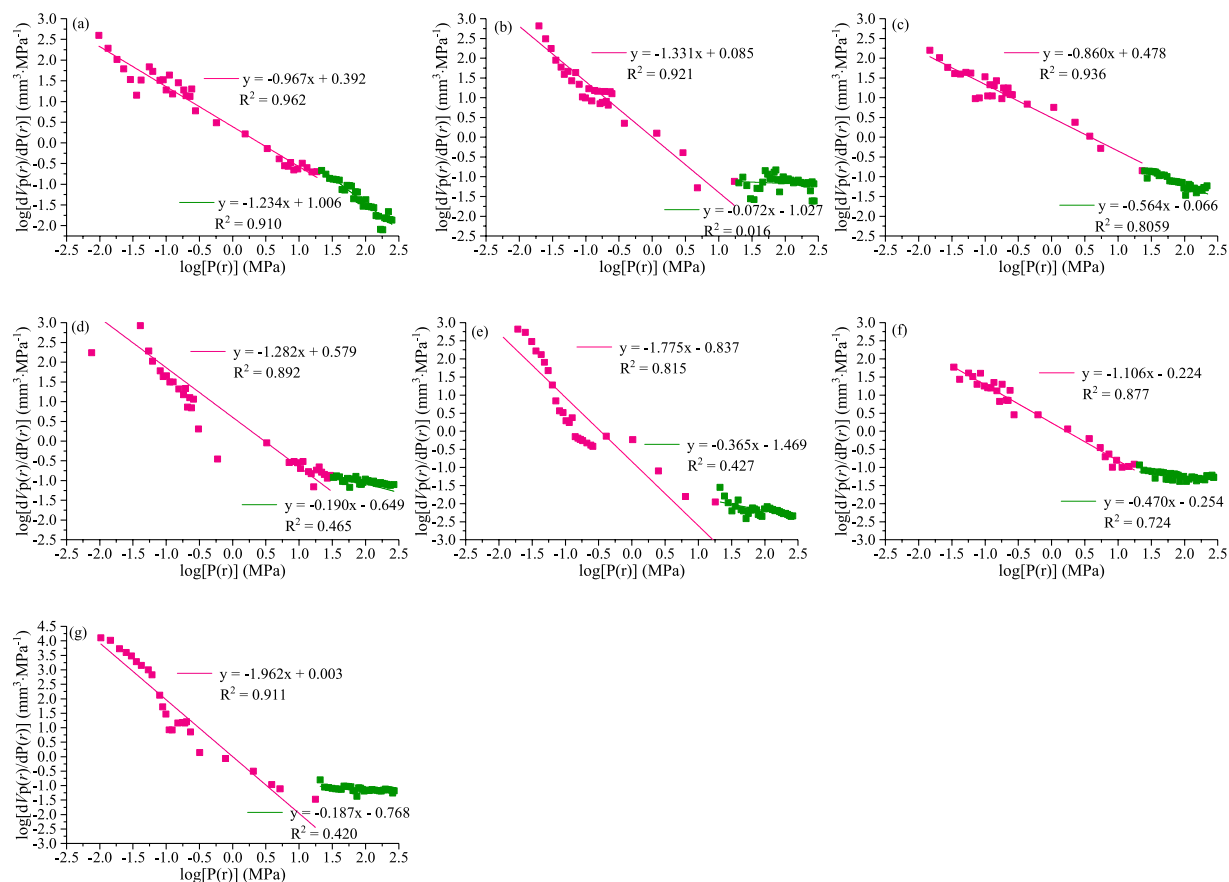


Figure 4. Fractal dimensions D_{s2} and D_{s1} for different coal ranks based on the sponge model.

0.82–0.96), whereas D_{g2} has a wide range of R^2 (0.02–0.91). The values of D_{g1} are widely distributed (2.04–3.14, 2.67 in average), indicating significant differences in the discontinuities and roughness of different coal ranks. However, the value of D_{g2} are 2.77–3.93 (3.56 in average). Most of the values of D_{g2} are close to 3, indicating that the surface is extremely rough and the pore structure is irregular⁵⁵. The values of D_{g1} in S_1 and S_3 are >3 , and all the values of D_{g2} are more than 3, except for sample S_1 . For the fractal dimensions of D_{g1} , numerous explanations have been provided in previous research^{28,44}.

Multifractal analysis. The spectrum curves of $\log\chi(q, \epsilon)$ versus $\log\epsilon$ show a linear relationship (correlation index R^2 , 0.78–1), indicating a multifractal distribution of the pore sizes⁵⁶ (Fig. 5). The spectrum curves of the generalized dimensions, $D(q)$, versus q present a sigma-shaped curve and follow a monotone decreasing function of q (Fig. 6a). The characteristic parameters of $D(q)$, dimensions D_0 , D_1 , and D_2 , Hurst exponent H ($2H = D_2 + 1$),

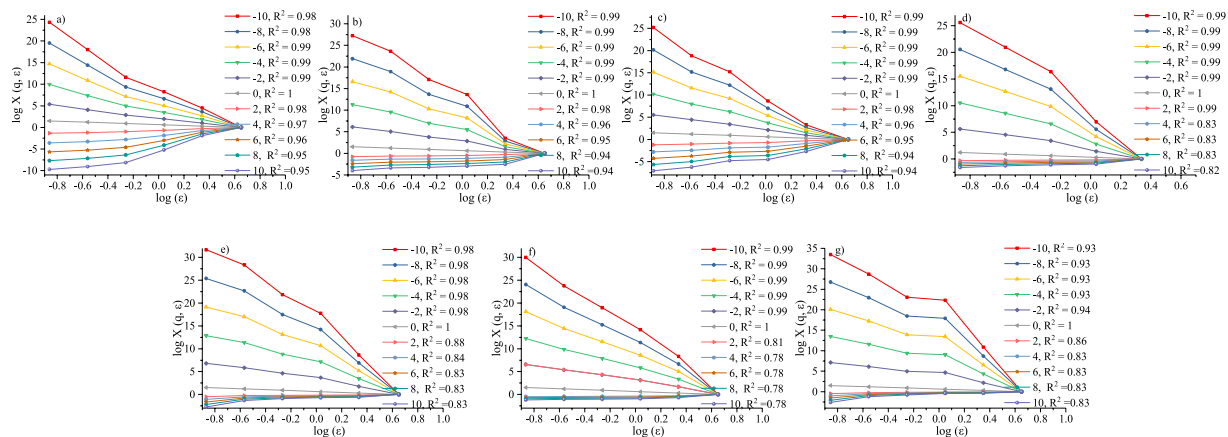


Figure 5. Partition function, $\log\chi(q, \epsilon)$, box size, and $\log\epsilon$ for the pore size distributions.

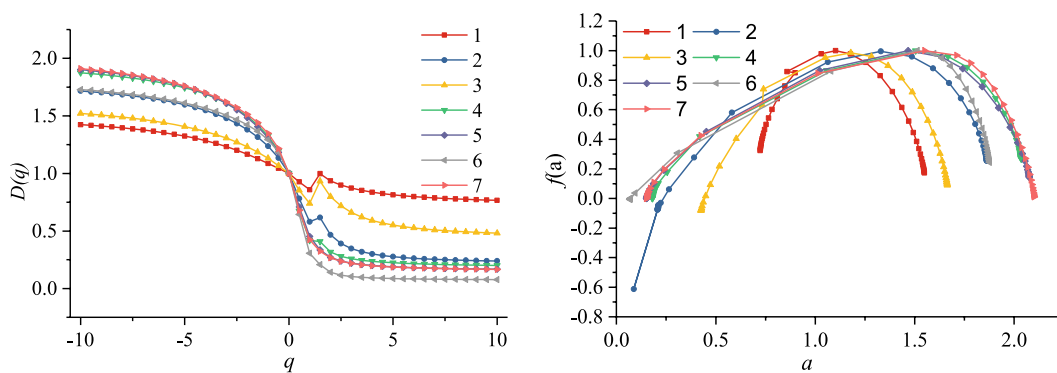


Figure 6. (a) Curve of D_q versus $q(-10, 10)$ of the coal samples and (b) the multifractal singularity spectra of the coal samples.

Samples	D_0	D_1	D_2	H	D_{10}	D_{-10}	$D_{-10}-D_{10}$	D_0-D_{10}	$D_{-10}-D_0$	a_0
S_1	1	0.86	0.94	0.97	0.77	1.42	0.66	0.23	0.42	1.10
S_2	1	0.58	0.47	0.73	0.24	1.72	1.48	0.76	0.72	1.33
S_3	1	0.74	0.80	0.90	0.48	1.52	1.04	0.52	0.52	1.18
S_4	1	0.42	0.32	0.66	0.20	1.88	1.67	0.80	0.88	1.51
S_5	1	0.45	0.27	0.64	0.17	1.90	1.73	0.83	0.90	1.47
S_6	1	0.31	0.14	0.57	0.08	1.73	1.65	0.92	0.73	1.51
S_7	1	0.42	0.27	0.63	0.17	1.91	1.74	0.83	0.91	1.55

Table 3. Multifractal parameters from the generalized dimension spectrum. D_0 , the capacity dimension; D_1 , the entropy dimension; D_2 , the correlation dimension; H, Hurst exponent; D_{10} and D_{-10} are the generalized dimensions responding to $q = 10$ and $q = -10$, respectively.

width $D_{-10}-D_{10}$ of $D(q)$ spectrum, right side width D_0-D_{10} , and left side width $D_{-10}-D_0$ (Table 3), reflect the inner variability and heterogeneity of the porosity and pore size distribution^{45,51,57}. Samples S_1 and S_6 have the highest and lowest values of D_1 , D_2 , H, and D_{10} , respectively. The values of D_1 widely range from 0.31 to 0.86.

Propensity to spontaneous combustion. Fig. 7 displays the results of the CPT measurements performed on the medium-rank and high-rank coals. The CPT ranges from 146.1 to 182.2 °C. Samples S_7 (anthracite B) and S_1 (bituminous D) have the highest CPT (182.2 °C) and the lowest CPT (146.1 °C), respectively. The CPT increases with increasing coal rank.

Discussion

Effect of evolution of coal petrology on spontaneous combustion. Relationship between Rr and CPT and pore structure. Coal rank has a significant influence on the propensity of coal to spontaneous combustion (Fig. 8a). A good linear relationship ($R^2=0.74$) is exhibited between the coal rank and the CPT. The CPT increases with increasing coal rank, which is in agreement with previous studies⁵⁸⁻⁶¹. However, sample S_6 exhibits

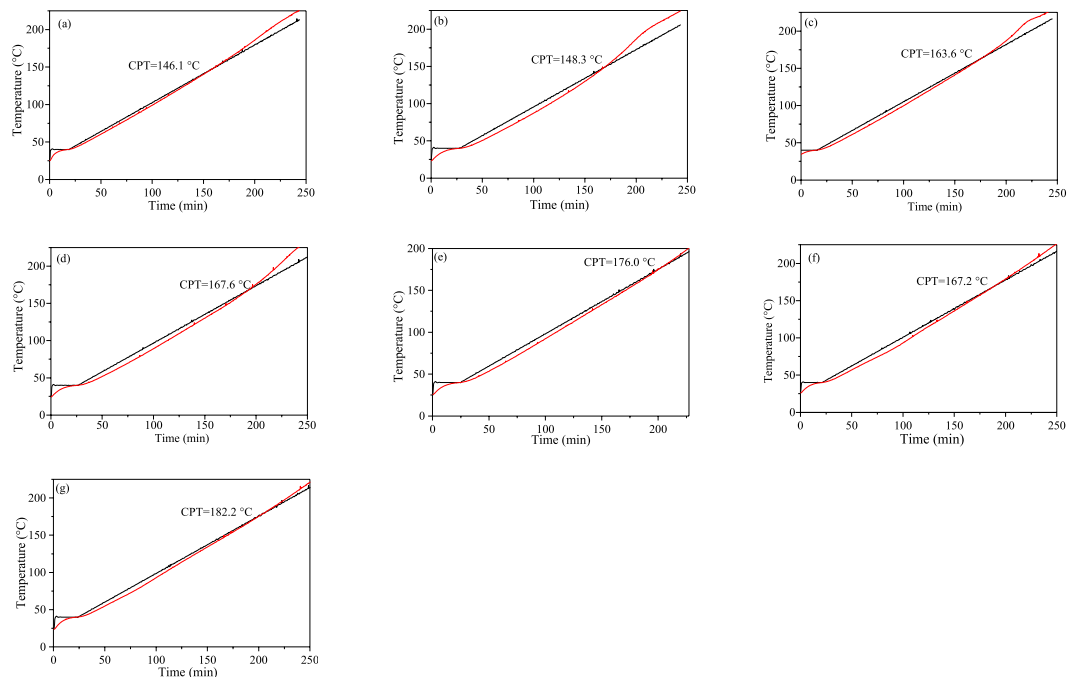


Figure 7. CPT for the different coal ranks.

a deviation compared to the other samples. If this sample data are removed, a better linear ($R^2 = 0.89$) is obtained, as shown in Fig. 8a. For the high-rank sample S_6 , the low CPT may be attributed to its different chemical (low fixed carbon) and physical (low total pore volume) structure, which is in agreement with previous studies^{11,12}. Coal rank is an important index of coalification, influencing the structures of coal pores and fractures. When the vitrinite random reflectance ranges from 0.58 to 2.13% (Fig. 8b), the mean volume increases with the coal rank, but the mean surface area decreases. When the vitrinite random reflectance ranges from 2.13 to 3.43% (Fig. 8c), the mean volume decreases with coal rank, but the mean surface area increases with coal rank. When the coalification ranges from 0.5 to 2.1%, the aromatic structures including non-protonated aromatic carbons (fa^N), nuclear magneton resonance (NMR) aromaticity (fa^i), and aromatic carbon ratio (fa) increase linearly, whereas the aliphatic structures decreases linearly⁶². Because the volume is mainly affected by the aliphatic parts of the chemical structure, it may be the cause of the displayed mean volume and surface area trends^{63,64}.

Relationship between coal composition and CPT. There exists a non-linear relationship between the ash content and the CPT, which shows an inverted ‘U-shape’ (Fig. 9a). The CPT is a weakly correlated to the total sulfur content (Fig. 9b). The coal components (volatile matter, oxygen contents, and fixed carbon contents) display good relationships with the CPT (Fig. 9c–e), which is in agreement with Zonguldak coals⁶⁵. The CPT decreases with increasing volatile matter, illustrating that low volatile matter is prone to spontaneous combustion. It may be that low volatile matter content can increase the difficulty of ignition and result in an unstable combustion flame^{66,67}. The relationship between the oxygen content and the CPT also shows the same trend as the volatile matter, indicating that coals with high oxygen content have a high tendency to chemically bind moisture, thereby rendering the surface, highly susceptible to autogenous heating⁶⁸. The CPT increases linearly with the fixed carbon content, indicating that less fixed carbon content is prone to spontaneous combustion. This is because a small amount of fixed carbon in coal requires a low activation energy to initiate combustion⁶⁹.

Relationship between pore structure and CPT. *Distribution of pore structure and CPT.* It is necessary to discuss the relationship between the pore size and the CPT. Fig. 10a shows a weak negative linear correlation between the cumulative pore volume V_1 and the CPT, which shows an inverted ‘U-shape’. Cumulative pore volume V_2 shows a ‘U-shape’ (Fig. 10b). The relationship between cumulative pore volume V_3 and the CPT exhibit a linear positive correlation (Fig. 10c). Moreover, the total pore volume also has a good linear positive correlation with the CPT (R^2 up to 0.71), indicating that pore sizes of more than 10000 nm (V_3) play a main role in coal spontaneous combustion (Fig. 10d). Specifically, the pore structure is a dominant factor causing the coal spontaneous combustion, particularly in pore sizes more than 10000 nm.

Relationship between coal rank, pore volume, and multifractal parameters. The entropy dimension (D_1) reveals the concentration degree of the porosity distribution⁴⁸. The values of D_1 are less than or equal to D_0 . When D_1 is close to D_0 , the porosity is evenly distributed. Otherwise, most particles are concentrated in a small area and appear as a high peak on the graph⁷⁰. Among of all sample S_1 has the highest homogeneous pore size distribution. D_1 decreases with increasing coal rank (Fig. 11a), indicating an increase in the heterogeneity. However, the fitting results do not a high correlation.

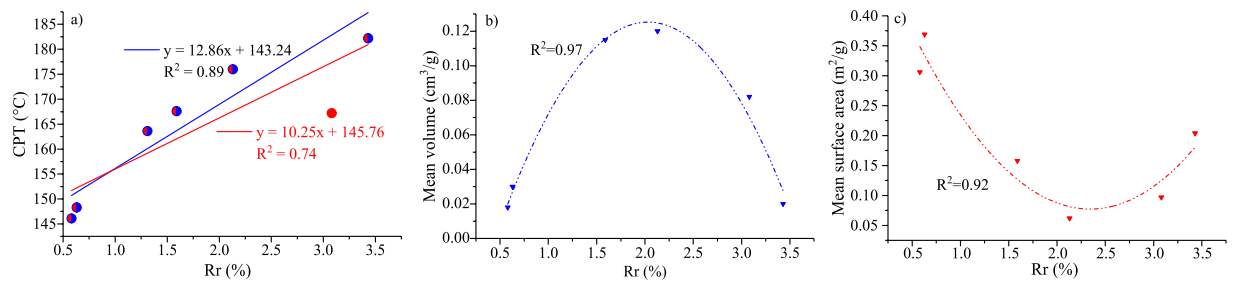


Figure 8. (a) Relationship between the vitrinite random reflectance (Rr, %) and (a) the CPT, (b) mean volume (cm³/g), and (c) mean surface area (m²/g).

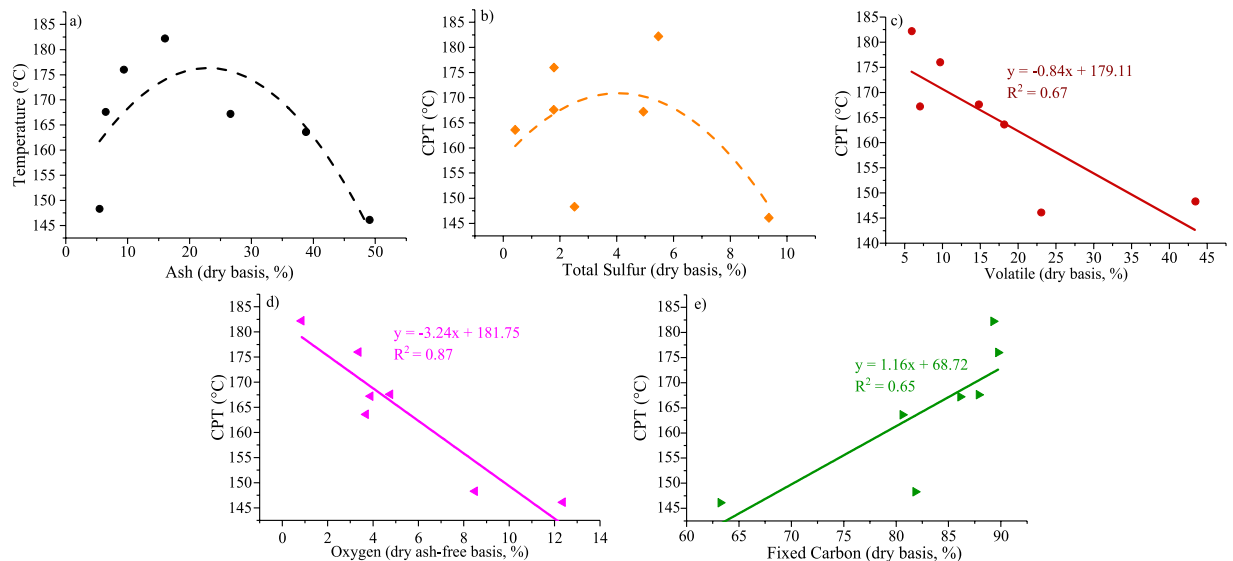


Figure 9. Relationship between the CPT and the (a) ash yield, (b) total sulfur content, (c) volatile matter, (d) oxygen content, and (e) fixed carbon content.

This may be owing to joint action factors such as maceral content, ash content, volatile matter, and tectonic deformations. The relationship between cumulative pore volume V_1 and D_1 is not clear, indicating that V_1 (adsorption-pores) may have little influence on the entropy dimension (D_1) (Fig. 11b). The D_1 increases with increasing cumulative pore volume V_2 (Fig. 11c) and decreases with increasing cumulative pore volume V_3 (Fig. 11d), indicating that seepage-pores have an important influence on the entropy dimension (D_1).

The Hurst exponent (H) is used to quantify the degree of correlation on the logarithmic scale⁷¹. If $H > 0.5$, the increments are correlated⁷¹. The Hurst exponent of all the samples exceeds more than 0.5 (H , 0.57–0.97), indicating the increments are correlated (Table 3). The Hurst exponents of samples S_1 and S_3 are 0.97 and 0.90, respectively, which two data are close to the 1, reflecting the presence of strong persistence or positive autocorrelations^{49,71}. The Hurst exponent has the same characteristics as entropy dimension (Fig. 12).

The width of $D(q)$ reflects the heterogeneity in the porosity distribution^{45,51}. Sample S_7 has the highest $D_{-10} - D_{10}$ (the widest spectrum), indicating the highest heterogeneity over the entire pore size range among all the coal samples^{45,51}. In contrast, sample S_1 has the lowest $D_{-10} - D_{10}$ (the narrowest spectrum), reflecting the lowest heterogeneity in the porosity distribution over the entire pore size range^{45,51}. Samples S_2 and S_6 have wider right side $D(q)$ spectra than the left side $D(q)$ spectra, indicating high dominance of the high porosity concentrations⁴⁵. However, samples S_{13} , S_{43} , S_{53} , and S_7 have wider left side $D(q)$ spectra than right side $D(q)$ spectra, indicating a small porosity concentration⁴⁵. The widths of the $D(q)$ ($D_{-10} - D_{10}$, $D_0 - D_{10}$, and $D_{-10} - D_0$) spectra increase with increasing vitrinite random reflectance (Rr, %) (Fig. 13a). The relationship between cumulative pore volume V_1 and width of the $D(q)$ spectra is not clear, indicating that V_1 (adsorption-pores) may have little influence on the width of the $D(q)$ spectra (Fig. 13b). The width of the $D(q)$ spectra logarithmically decreases with increasing cumulative pore volume V_2 (Fig. 13c) and logarithmically increases with increasing cumulative pore volume V_3 (Fig. 13d). The above indicates that seepage-pores have an important influence on the heterogeneity in the porosity distribution.

Fractal dimensions and CPT. 1) Multiscale fractal dimensions and CPT. The relationships between the fractal dimensions of the Sierpinski model (D_{s1} and D_{s2}) and the CPT are shown in Fig. 11. The relationship between D_{s1} and the CPT can be divided into two distinct stages (Fig. 14a). In the first stage, the CPT first decreases,

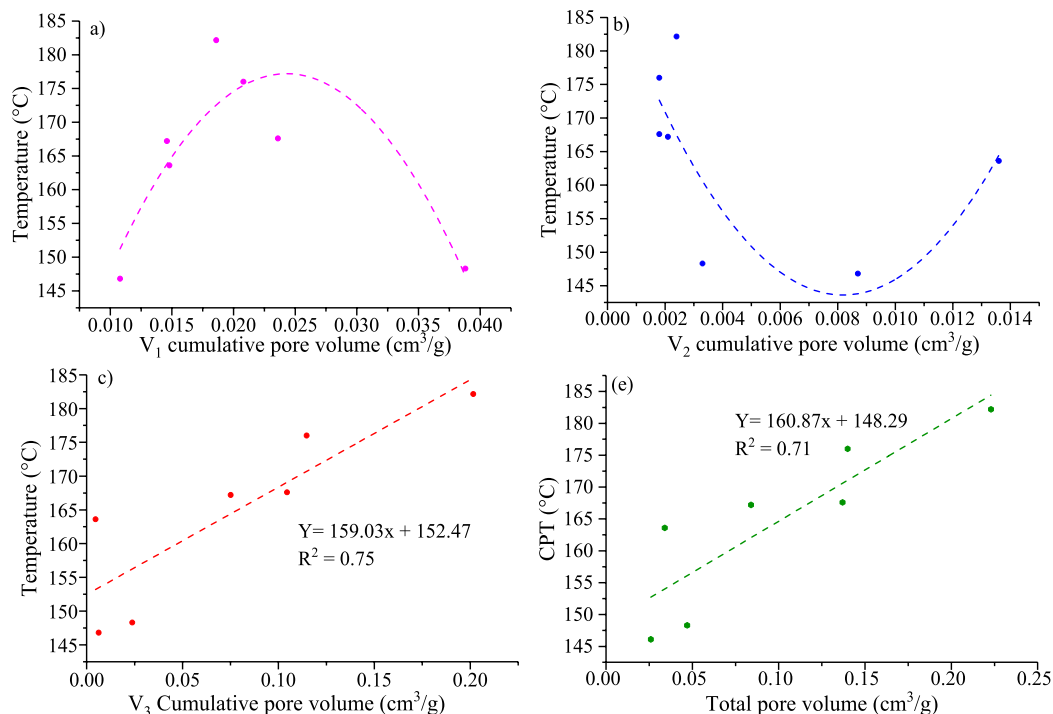


Figure 10. Relationship between CPT and the (a–d) pore volume distribution, and (e) total pore volume.

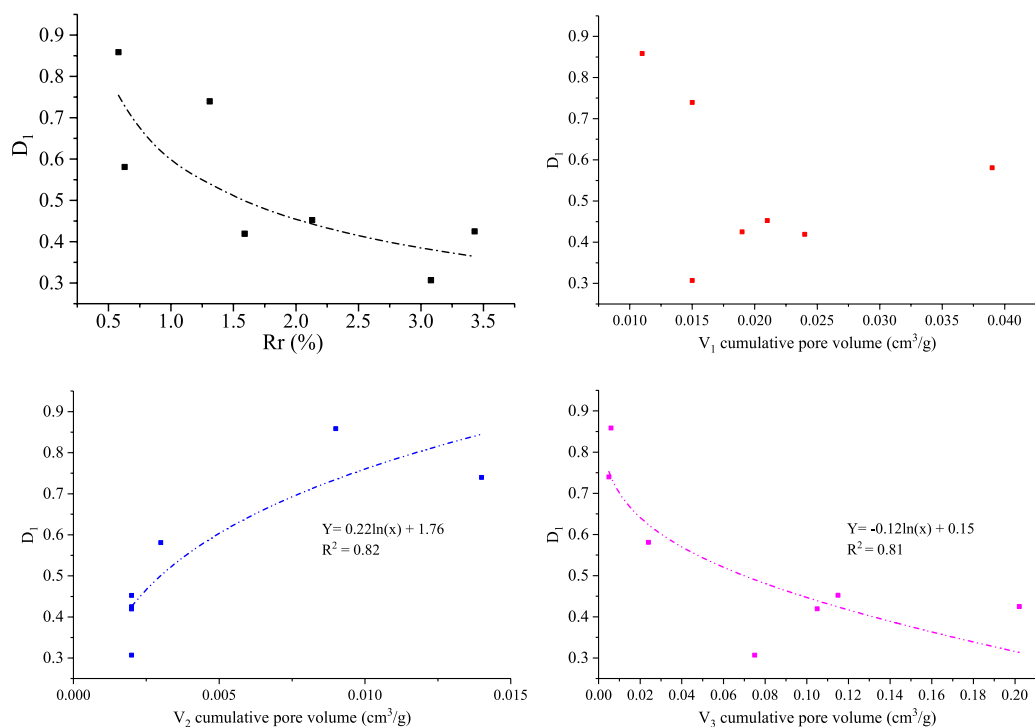


Figure 11. Relationship between D_1 and the (a) vitrinite random reflectance (R_r , %) and (b–d) Cumulative pore volume (V_1 , V_2 and V_3).

and then increases with increasing D_{s1} (when $D_{s1} > 2.8$). With increasing D_{s2} , the CPT decreases first (from 2.75 to 2.85), and then increases (Fig. 14b). Fractal dimensions D_{g1} of samples S_1 and S_3 are >3 ; however, most of the fractal dimensions, D_{g2} , are >3 , except of sample S_1 . Fractal dimensions D_{g1} and D_{g2} are >3 , which are not suitable to characterize the pore heterogeneity^{72,73}. To prevent the interference of abnormal points, the data for fractal dimensions >3 were removed³⁷. The relationship between D_{g1} and the CPT (without samples S_1 and S_2)

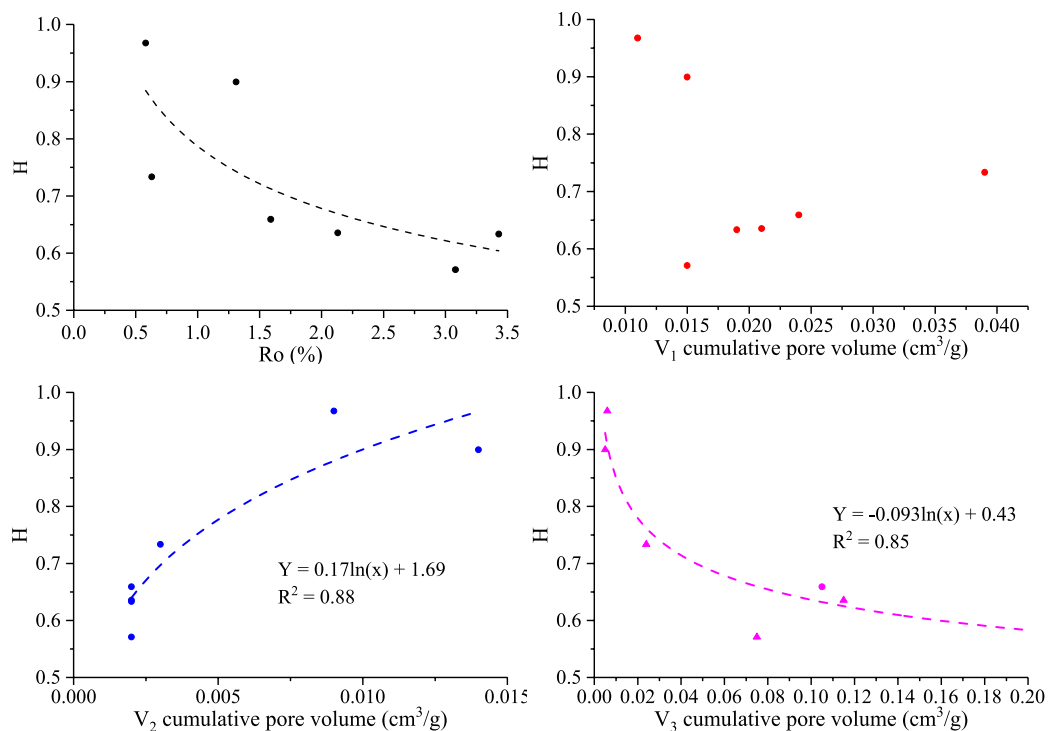


Figure 12. Relationship between the Hurst exponent (H) and the (a) vitrinite random reflectance (Rr , %) and (b–d) Cumulative pore volumes (V_1 , V_2 , and V_3).

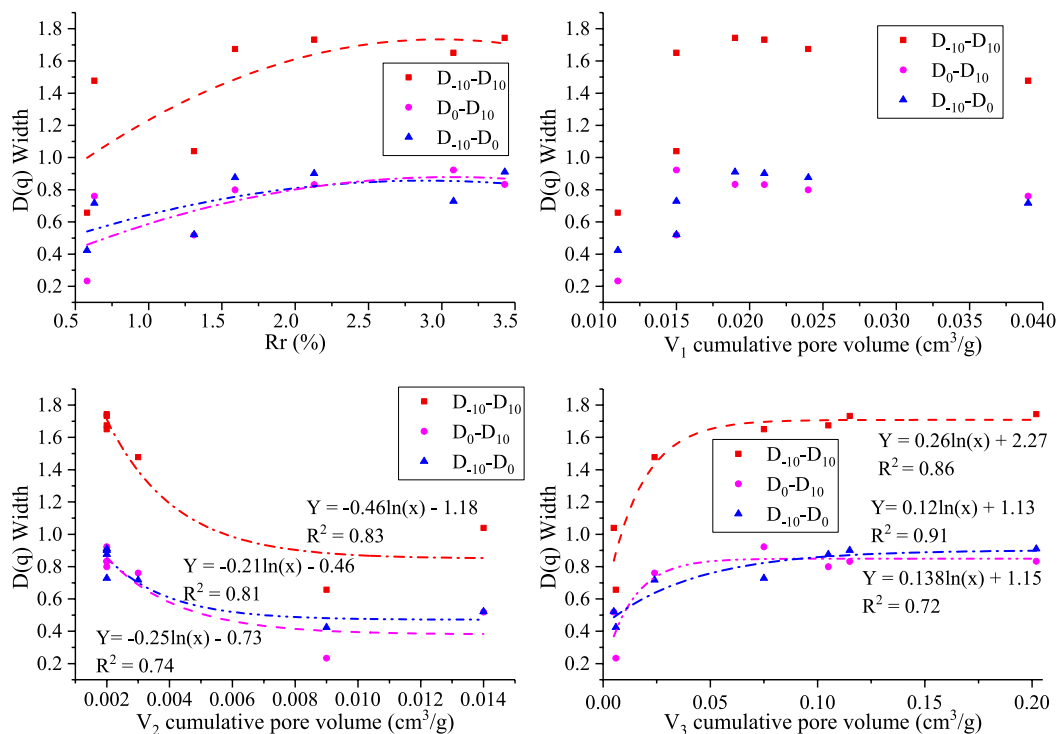


Figure 13. Width of the $D(q)$ spectra ($D_{-10}-D_{10}$, D_0-D_{10} , and $D_{-10}-D_0$) and the (a) vitrinite random reflectance (Rr , %) and (b–d) cumulative pore volumes (V_1 , V_2 , and V_3).

displays a similar trend as that between D_s and the CPT obtained from the Sierpinski model (Fig. 14c). When D_{g1} ranges from 2 to 2.6, the CPT decreased significantly. When D_{g1} becomes larger than 2.6, the CPT increases with increasing D_{g1} . The above results demonstrate that the heterogeneities obtained from the Sierpinski and Sponge

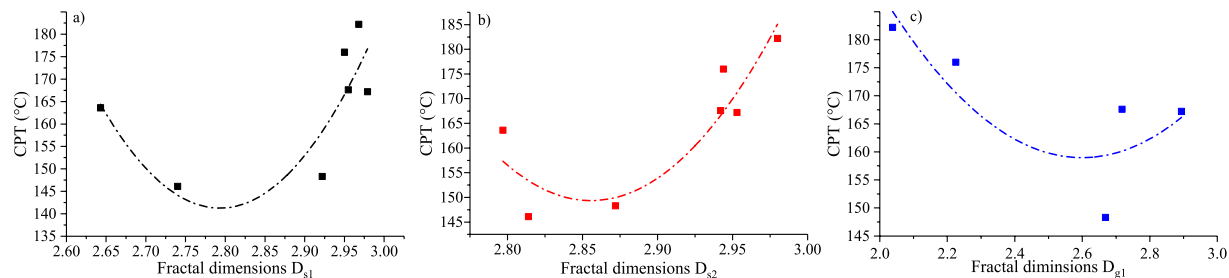


Figure 14. Relationship between the CPT and (a) fractal dimensions D_{s1} , (b) fractal dimensions D_{s2} , and (c) fractal dimensions D_{g1} .

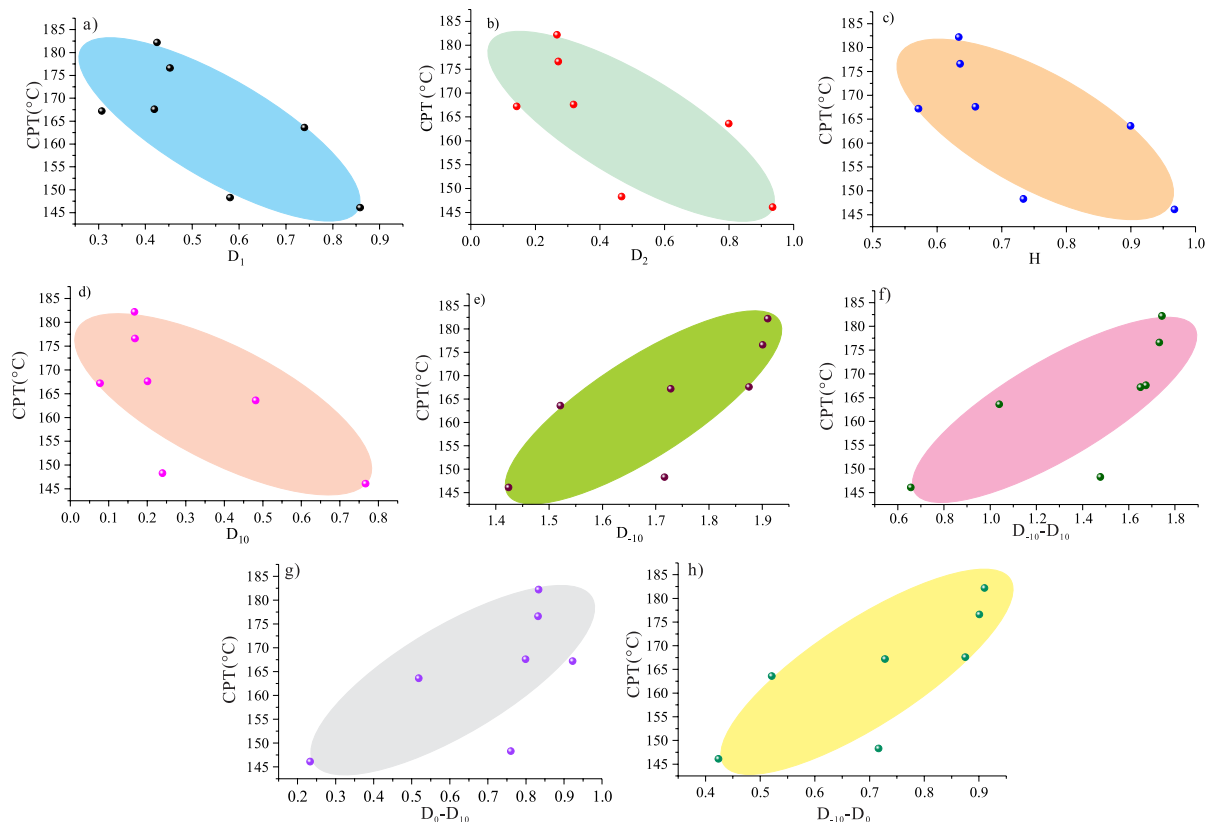


Figure 15. Relationship between the CPT and the multifractal parameters.

models do not present linear relationship with the tendency of coal to spontaneous combustion. However, a high heterogeneity (fractal dimensions >2.8) is associated with a low tendency of spontaneous combustion.

2) Multifractal fractal dimensions and CPT. Fig. 15 displays the correlation between the multifractal fractal dimensions ($D_1, D_2, H, D_{10}, D_{-10}, D_{-10}-D_{10}, D_{-10}-D_0, D_0-D_{10}$) and the CPT. It can be found that the CPT decreases with increasing parameters D_1, D_2, H , and D_{10} , suggesting that a high degree of the distribution of the porosity quantifies the degree of correlation on the logarithmic scale. Further, the heterogeneity in the porosity distribution decreases the tendency of coal spontaneous combustion. The correlation between $D_{-10}, D_{-10}-D_{10}, D_{-10}-D_0$, and D_0-D_{10} and the CPT is clear in that the CPT gradually increases with increasing $D_{-10}-D_{10}, D_{-10}-D_0$, and D_0-D_{10} . This indicates that the complexity of the local characterization pore structure decrease the spontaneous combustion propensity.

Conclusions

The conventional analysis and CPT measurements were conducted to obtain the properties of coal petrology and spontaneous combustion. Coal rank, volatile matter, oxygen content, and fixed carbon content were found to play important roles in spontaneous combustion.

The pore structure properties obtained from HPMTI provided a direct measurement of coal physical properties. The cumulative pore volume of $V_3 (>10000 \text{ nm})$ and total pore volume have positive correlation with CPT.

Multiscale and multifractal analyses were conducted to evaluate the pore size distribution. From the multi-scale analysis, the relationship between the fractal dimensions (D_{s1} , D_{s2} , and D_{g1}) and the CPT basically displayed a “U-shaped” tendency, with the minimum occurring at 2.6–2.85. Based on the multifractal analysis, a high degree of porosity distribution, quantified the degree of correlation on the logarithmic scale. Furthermore, the heterogeneity in the porosity distribution decreased the tendency of coal spontaneous combustion; therefore, a more complex local characterization pore structure lowered the spontaneous combustion propensity.

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Author contributions

Zhang Yu: Data processing and paper writing. Zhang Xueqing: Samples preparation. Yang Wen: Samples collection. Xin Haihui: Experimental test. Hu Sherong: Funding and ideas. Song Yu: Method and paper revision.

Competing interests

The authors declare no competing interests.

Additional information

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