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Pitting Corrosion and Microstructure of J55 Carbon Steel Exposed to CO2/Crude Oil/Brine Solution under 2–15 MPa at 30–80 ◦**C**

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Abstract: This study aimed to evaluate the corrosion properties of J55 carbon steel immersed in $CO₂/crude oil/brine mixtures present in the wellbores of CO₂-flooded production wells. The main$ corroded position of wellbore was determined and wellbore corrosion law was provided. Corrosion tests were performed in 30% crude oil/brine solution under the simulated temperature (30–80 ◦C) and pressure (2–15 MPa) conditions of different well depths (0–1500 m). The corrosion behavior of J55 carbon steel was evaluated through weight-loss measurements and surface analytical techniques, including scanning electron microscopy, energy dispersive spectrometer, X-ray diffraction analysis, and optical digital microscopy. Corrosion rate initially increased and then decreased with increasing well depth, which reached the maximum value of 1050 m. At this well depth, pressure and temperature reached 11 MPa and 65 °C, respectively. Under these conditions, FeCO₃ and CaCO₃ localized on sample surfaces. Microscopy was performed to investigate corrosion depth distribution on the surfaces of the samples.

Keywords: CO₂ corrosion; J55 carbon steel; CO₂/crude oil/brine mixtures; corrosion depth distribution

1. Introduction

 $CO₂$ is internationally recognized as a major greenhouse gas that accounts for approximately 65% of the total greenhouse gas emissions [\[1](#page-11-0)[–3\]](#page-11-1). All countries currently attach considerable importance to environmental issues, particularly global warming caused by $CO₂$. $CO₂$ is used as an oil-flooding agent worldwide because it can effectively reduce crude oil viscosity and residual oil saturation, dissolve gum in reservoirs, and increase permeability and crude oil recovery rate $[4,5]$ $[4,5]$. CO₂ flooding can reduce CO2-associated air pollution and greenhouse effects. Nevertheless, oil pipe failure caused by CO² corrosion has become a commonly encountered problem in oilfields and results in great economic losses and safety hazards while severely restricting the development of $CO₂$ flooding technology [\[4–](#page-11-2)[6\]](#page-11-4).

CO² corrosion and its control in oil casing and surface transmission pipelines have been important topics in the field of oil and gas exploration. Related studies have focused on the influence of environmental and material factors on corrosion behavior [\[7–](#page-11-5)[16\]](#page-11-6). For example, the corrosive medium and environment have been identified as the deciding factors of corrosion rate and morphology. The influence of temperature and pressure on corrosion rate is mainly reflected by the changes that they induce in the protectiveness of the corrosion product layer [\[7–](#page-11-5)[18\]](#page-12-0). These changes, in turn, lead to changes in the corrosion rate. With increasing temperature, the $CO₂$ corrosion rate of carbon steel negligibly changes with pressure $(4, 8, \text{ and } 12 \text{ MPa})$. The $CO₂$ corrosion rate of carbon steel increases with the water cut of water/crude oil mixtures [\[15](#page-11-13)[–18\]](#page-12-0). Nevertheless, efficient anticorrosion measures cannot be established on the basis of the results of preliminary studies that have largely focused on the influence of a single factor on $CO₂$ corrosion and of the limited works that have investigated the laws governing CO_2 corrosion in the wellbores of CO_2 -flooded wells. Although studies on aqueous CO_2 corrosion in crude oil/brine mixtures have been performed, comprehensive studies on $CO₂$ corrosion in CO_2/c rude oil/brine mixtures at different well depths remain unavailable.

This work explored the laws governing $CO₂$ corrosion at different depths of $CO₂$ -flooded wells. Specifically, this study investigated the common effects of temperature and pressure on $CO₂$ corrosion in a CO_2/c rude oil/brine environment. The average corrosion rate was quantified through the weight-loss method. The surface morphologies of corrosion samples were observed through scanning electron microscopy (SEM). The compositions of corroded surfaces were analyzed through energy dispersive Spectroscopy (EDS) and X-ray diffraction (XRD) analyses. Corrosion depth distribution was analyzed through microscopy, 42 percent of the exposed surface area was studied, which can accurately reflect the surface conditions of the samples after corrosion.

2. Materials and Methods

2.1. Experimental Materials

J55 carbon steel was processed into rectangular samples (50 mm \times 10 mm \times 3 mm, φ = 6 mm) for the weight-loss test and surface analysis. The chemical composition of J55 carbon steel is shown in Table [1.](#page-1-0) The samples were placed in acetone to remove surface oil and then immersed in ethanol for 5 min for degreasing and dehydration. The samples were collected, dried with cold air, packed in filter paper, and placed in a dryer for 4–7 h. The sizes and weights of the samples were measured within an accuracy of 0.1 mg.

Elements (wt.%)								
		Mn			Сr	Ni	('11	Fe
		$0.34 - 0.39$ $0.2 - 0.35$ $1.25 - 1.5$ < 0.020		< 0.015	≤ 0.15	< 0.20	< 0.20	Bal.

Table 1. Chemical composition of type J55 carbon steel (wt.%).

The corrosive medium comprised crude oil and brine. Crude oil was collected from the Chang-8 oil reservoir, a certain block in Changqing Oilfield, and the composition of simulated brine was based on the composition of the water produced in the Chang-8 oil reservoir. The compositions of crude oil and brine are shown in Tables [2](#page-1-1) and [3,](#page-2-0) respectively.

Property	Units	Value
NaCl	$g \cdot L^{-1}$ g $\cdot L^{-1}$	18.5028
CaCl ₂		13.7338
MgCl ₂		0.5897
Na ₂ SO ₄		0.2440
NaHCO ₃		0.0631
Salinity	$g \cdot L^{-1}$ g $\cdot L^{-1}$ g $\cdot L^{-1}$ g $\cdot L^{-1}$	33.0000

Table 3. Properties of simulated brine. Table 3. Properties of simulated brine

2.2. Weight-Loss Corrosion Test 2.2. Weight-Loss Corrosion Test

The corrosion test was performed in accordance with the weight-loss method with a PARR-4578 The corrosion test was performed in accordance with the weight-loss method with a autoclave. The schematic of the test is shown in Figure [1.](#page-2-1) 1 L the mixture of crude oil and brine ($v:v = 3:7$) was added to the autoclave and purged with a small amount of N_2 for 120 min under 0.5 MPa to remove dissolved O_2 . Then, the mixture was subjected to 60 min of injection with high-purity CO_2 under 1 MPa to remove N_2 [\[19\]](#page-12-1). Finally, autoclave temperature was increased to the test temperature, and autoclave pressure was increased to the test pressure with high-purity $CO₂$. The test conditions were maintained for 2 days at a running speed of $0.5\,\mathrm{m\cdot s^{-1}}$ (200 r \cdot min $^{-1}$).

Figure 1. Flow chart of the evaluation system for steel corrosion rate (mass-loss method). **Figure 1.** Flow chart of the evaluation system for steel corrosion rate (mass-loss method).

accordance with the *ASTM G1-03 Standard Practice for Preparing, Cleaning, and Evaluating Corrosion Test Specimens* [\[20\]](#page-12-2). The samples were immediately rinsed with distilled water. Acetone was used to remove crude oil from the surfaces of the samples after corrosion induction. Then, the samples were immersed in acid cleaning solution (500 mL of HCl and 3.5 g of hexamethylenamine brought to volume with water to 1000 mL) for 10 min. At the same time, the corrosion products were removed from the surfaces of the samples, and the samples were collected from the acid cleaning solution. The acid cleaning solution was thoroughly rinsed off from the surfaces of the samples with distilled water. The samples were then twise immersed in other of gleaning and debydration, sellected water. The samples were then twice immersed in ethanol for cleaning and dehydration, collected, water. The samples were then twice immersed in ethanol for cleaning and dehydration, collected,The corrosion rate of the corroded steel was determined through the mass-loss method in

placed on filter paper, dried with cold air, packed in filter paper, and placed in the dryer for 4–7 h. Finally, the samples were weighed to within an accuracy of 0.1 mg. Corrosion rate was calculated using the following formula: Finally, the samples were weighed to within an accuracy of 0.1 mg. Corrosion rate was calculated m aced on their paper, dried θ 4.76

$$
r_{corr} = \frac{8.76 \times 10^4 \times (m - m_t)}{S \times t \times \rho}
$$
 (1)

where r_{corr} is average corrosion rate (mm·y⁻¹); *m* and m_t are the weights of the test sheet before and after the experiment, respectively (g); *S* is the area of the whole surface in contact with the solution after the experiment, respectively (g); *S* is the area of the whole surface in contact with the solution (cm^2) ; ρ is the density of the tested steel $(\text{g}\cdot\text{cm}^{-3})$; and t is the duration of immersion (h). Each test was performed with three parallel samples. The mean corrosion rate error was calculated on the basis of performed with three parallel samples. The mean corrosion rate error was calculated on the basis of the results for the three parallel samples. the results for the three parallel samples.

2.3. Microstructure Observation 2.3. Microstructure Observation

After corrosion induction, samples were extracted from the autoclave and rinsed with distilled After corrosion induction, samples were extracted from the autoclave and rinsed with distilled water and acetone. The surface microstructures of the corrosion product layers on the surfaces of water and acetone. The surface microstructures of the corrosion product layers on the surfaces of corroded samples were analyzed through SEM with FEI Quantu 600F microscope (Hillsboro, OR, corroded samples were analyzed through SEM with FEI Quantu 600F microscope (Hillsboro, OR, USA). The elemental compositions of the corrosion product layers were determined with OXFORD USA). The elemental compositions of the corrosion product layers were determined with OXFORD INCA energy 350. The compositions of the corrosion product characterized were determined through INCA energy 350. The compositions of the corrosion product characterized were determined XRD by using Bruker D8 XRD (Billerica, MA, USA).

2.4. Statistics of Corrosion Depth Distribution 2.4. Statistics of Corrosion Depth Distribution

The surface depths of the corroded samples cleaned using acid cleaning solution were visualized The surface depths of the corroded samples cleaned using acid cleaning solution were using an OLYMPUS DSX500 optical digital microscope (Tokyo, Japan). Sample surfaces were subjected to grand horizon 3D image capture under bright-field mode with the adjacent visual threshold splicing mode. As shown in Figure [2,](#page-3-0) images for the analysis of surface corrosion morphology were acquired at eight observation points on both sides of the same sample through nine-field splicing under 200 \times magnification and 10% coincidence rate. The display heights of the 3D images were adjusted to the maximum pitting height to ensure that the different ranges of corrosion depths in the same set of images can be represented by different colors. Identical corrosion depth ranges were represented by the same color. Finally, the 3D images were converted into the contour diagrams of corrosion depth distribution. diagrams of corrosion depth distribution.

Figure 2. Schematic of image acquisition. **Figure 2.** Schematic of image acquisition.

3. Results and Discussion

3.1. Corrosion Law of the Deepening Well Materials 2018, 11, x For PEER REVIEW 5 of 13, x FOR PEER REVIEW 5 of 13, x FOR PEER REVIEW 5 of 13, x FOR PEE

The CO₂-flooded well in Chang-8 Oil Reservoir of a certain block in Changqing Oilfield was taken as an example. The experimental well had a depth of 1550 m, a well-head temperature of 30 ℃, pressure of 2 MPa, and well-bottom temperature of 82 °C. Temperature and pressure distributions as a function of well depth are shown in Figure 3. Tests were performed at 0 m (2 MPa, 30 °C), 240 m (4 MPa, 40 °C), 580 m (7 MPa, 50 °C), 800 m (9 MPa, 55 °C), 1050 m (11 MPa, 65 °C), and 1500 m (15 MPa, 80 °C). The liquid produced by the well had a water cut of 70%. The crude-oil composition of the liquid is shown in Table 2. The chemical composition of water produced by the well is shown in Table 3.

Figure 3. Distribution of temperature and pressure as a function of well depth.

the average corrosion rate and the maximum corrosion depth of the samples corroded at different well depths. The average corrosion rate initially increased and then decreased with well depth, except for 1050 m. The general equations for the anodic and cathodic reactions of $CO₂$ corrosion in deoxygenated solution are shown as Equations (2) and (3), respectively [\[9\]](#page-11-8). Figure [4](#page-5-0) shows the appearance of samples after the removal of corrosion scales. Figure 5 shows

$$
\text{Fe} \to \text{Fe}^{2+} + 2e^- \tag{2}
$$

$$
CO2 + H2O \rightarrow H2CO3
$$
\n(3)

the viscosity and protective effect of crude oil on the sample surfaces while intensifying mass transfer between the samples and corrosive medium and accelerating corrosion. Moreover, CaCO₃ and FeCO₃ deposits generated from the reaction between CO_3^2 and HCO_3^- and between Ca^{2+} and Fe^{2+} in →
the liquid gradually increased and inhibited corrosion development by forming a protective layer on the sample surfaces [\[8](#page-11-7)[,9\]](#page-11-8). The increase in CO₂ pressure reduced system pH and is conducive for the formation of the protective corrosion product layer [\[21\]](#page-12-3). Hence, corrosion acceleration and inhibitory effects simultaneously occurred in the system. At 580 m, the maximum corrosion depth sharply increased, and corrosion type shifted from uniform to local corrosion because the corrosive environment transformed from a $CO₂/c$ rude oil/brine environment to a subcritical $CO₂/c$ rude oil/brine environment [\[22\]](#page-12-4). The scaling ability of the solution decreased, the surface of the sample could not be completely covered by precipitates, and the progress of the anodic reaction could only $\frac{1}{2}$ be partially prevented [23]. These effects resulted in local corrosion. Therefore, the average corrosion be partially prevented [\[23\]](#page-12-5). These effects resulted in local corrosion. Therefore, the average corrosionTemperature and pressure increased as well depth increased. Increasing temperatures reduced

rate was low (1.7658 mm·year⁻¹), whereas the maximum corrosion depth was high (164.358 µm). Under increasing temperature and pressure, the corrosive environment transformed to the supercritical $CO₂/crude oil/brine environment and was dominated by corrosion. Hence, the maximum average$ corrosion rate was observed at the depth of 1050 m. The possible reduction in the base level of corrosion depth measurement may have reduced the maximum corrosion depths of the samples corroded at 1050 m. With the further increase in well depth, the pH of the corrosion system became almost constant, and the buildup of $CaCO₃$ and FeCO₃ deposits on the sample surfaces could suppress anodic dissolution [\[23\]](#page-12-5). The average corrosion rate decreased. Under dynamic conditions, however, the uneven coverage of the sample surfaces with $CaCO₃$ and FeCO₃ deposits resulted in however, the unevent coverage of the camp recurricts with CaCO3 and FeCO3 dependences. corrosion and increased the maximum corrosion depth of the samples corroded at 1500 $\,$ suppress anodic dissolution [23]. The average corrosion rate decreased. Under dynamic conditions, h wer, the uneven coverage of the sample surfaces with CaCO3 and FeCO3 deposits resulted in σ

Figure 4. Appearance of samples corroded at the well depths of (a) 0, (b) 240, (c) 580, (d) 800, (e) 1050, and (f) 1500 m after the removal of corrosion scales. and (**f**) 1500 m after the removal of corrosion scales. and (f) 1500 m after the removal of corrosion scales.

Figure 5. Average corrosion rate and maximum corrosion depth at different well depths.

3.2. Microstructures and Compositions of Corrosion Scales

The SEM images of the samples after corrosion at different temperatures and pressures are shown in Figure [6.](#page-6-0) The results for the spectral analysis of the corroded sample surfaces are shown in Table [4.](#page-7-0) Before CO_2 reached a supercritical state, the samples exhibited dense surface coatings that mainly $\frac{1}{10}$ consisted of FeC₃ and FeCO₃ and almost lacked CaCO₃ [\[14,](#page-11-12)[17,](#page-12-6)[19,](#page-12-1)[24\]](#page-12-7). The average corrosion rates and maximum corrosion depths of the samples were low. After CO_2 reached the supercritical state, loose surface coatings that mainly consisted of $FeCO₃$ and $CaCO₃$ formed and failed to provide effective surface protection to the samples $[14,19,24]$ $[14,19,24]$ $[14,19,24]$. Thus, the average corrosion rates and maximum corrosion depths of the samples increased. maximum corrosion depths of the samples increased.

Figure 6. SEM images (2000 \times) of the surface morphologies of J55 steel after corrosion at (a) 0, (b) 240, (c) 580, (d) 800, (e) 1050, and (f) 1500 m (**c**) 580, (**d**) 800, (**e**) 1050, and (**f**) 1500 m.

complex:

Well Depth/m Element	0	240	580	800	1050	1500
C K	25.04	35.63	45.36	17.94	19.14	9.92
O K	31.97	22.40	17.52	34.22	42.21	39.24
ClK	0.26				0.57	1.38
Ca K	1.10			4.09	10.51	4.70
Mn K	0.69		0.23		0.54	1.39
Fe K	38.89	41.97	36.89	36.47	26.78	42.08
Cu K	2.04			6.12		1.29
S K				1.16	0.26	
total content	100	100	100	100	100	100

Table 4. EDS results for the surfaces of J55 steel after corrosion (%).

Figure 7 shows the XRD spectra of the surface la[yer](#page-7-1)s of the corroded samples immersed in $CO₂/crude oil/brine mixtures under the given temperatures and pressures at different well depths.$ FeCO₃ is the main product of the CO₂ corrosion of carbon steel [\[7–](#page-11-5)[19](#page-12-1)[,25\]](#page-12-8). Similarly, the corrosion product layer that formed on samples immersed in CO_2/c rude oil/brine mixtures mainly comprised CaCO₃ and FeCO₃ complex salts. The specific composition of the product layer may be attributed to the isomorphous substitution of metal cations during CO₂ corrosion [\[23\]](#page-12-5). When [Fe²⁺] × [CO₃²⁻] in the medium exceeded the FeCO₃ solubility product K_{sp} (FeCO₃), that is, when the FeCO₃ supersaturation in the medium was $S = \{ [Fe^{2+}] \times [CO_3^{2-}] \} / \{K_{sp} (FeCO_3) \} > 1$, FeCO₃ was deposited on the surfaces of the samples. FeCO₃ deposition can be represented as follows [26]:

$$
\text{Fe}^{2+} + \text{CO}_3{}^{2-} \rightarrow \text{FeCO}_3 \text{ (s)}
$$
 (4)

Figure 7. XRD spectra of the surface layers of the corroded samples. **Figure 7.** XRD spectra of the surface layers of the corroded samples.

 Ca^{2+} in the solution replaced Fe²⁺ in the FeCO₃ crystal and ultimately formed the Fe(Ca)CO₃ complex:

$$
\text{Ca}^{2+} + \text{FeCO}_3 \text{ (s)} \rightarrow \text{Fe}^{2+} + \text{CaCO}_3 \text{ (s)} \tag{5}
$$

only bring at the well depth of 800 m. Figure 8a,b were obtained through optical digital microscopy of 800 m. Figure 8a,b were obtained through optical microscopy of 800 m. Figure 8a,b were obtained through optical micros under 200× magnification with nine-field splicing and 10% coincidence rate. Figure 8c shows the *3.3. Law of Corrosion Depth Distribution*

Figure 8 shows the results for the corrosion depth analysis of the sample corroded in 30% crude oil/brine at the well depth [of](#page-8-0) 800 m. Figure 8a,b were obtained through optical digital microscopy under $200 \times$ magnification with nine-field splicing and 10% coincidence rate. Figure [8c](#page-8-0) shows the contour diagram of the corrosion depth distribution of Figure [8a](#page-8-0), which was transformed from

Figure [8b](#page-8-0). The size of one contour diagram of corrosion depth distribution was 7612 μ m \times 7612 μ m, and the total image area observed was 57.94 mm², which accounted for 42% of the sample surface area. Therefore, this corrosion depth analysis method can accurately reflect the surface conditions of the samples after corrosion. Fig[ure](#page-8-1) 9 shows the contour diagrams of corrosion depth distribution at different observation positions after the removal of corrosion scales from the sample corroded in 30% crude oil/brine at the well depth of 800 m. 30% crude oil/brine at the well depth of 800 m. 30% crude oil/brine at the well depth of 800 m.

Figure 8. Corrosion depth analysis of the sample corroded in 30% crude oil/brine at the well depth of 800 m: (a) corrosion morphology, (b) 3D diagram of corrosion depth distribution, and (c) contour diagram of corrosion depth distribution.

Figure 9. Surface height distribution of the sample corroded in 30% crude oil/brine at the well depth of 800 m after the removal of corrosion scales ($100\times$).

The frequency density distribution of corrosion depth on the surface of the sample corroded at the well depth of 800 m is shown i[n Fi](#page-9-0)gure 10. The class intervals used to represent corrosion depth ranges i[n F](#page-8-1)igure 10 are the same as colors used to represent corrosion depth range[s in](#page-9-0) Figure 9. The plot of frequency density was bell-shaped with bilateral symmetry, wherein high values clustered in the center of the plot and low values clustered at both ends of the plot; these characteristics are indicative of typical Gaussian distribution [\[27,](#page-12-10)[28\]](#page-12-11). The frequency density distribution maps of samples corroded at different well depths sho[wn i](#page-9-1)n Figure 11 exhibit similar characteristi[cs o](#page-9-0)f Figure 10.

different well depths.

Figure 10. Frequency density distribution of corrosion scales. **Figure 10.** Frequency density distribution of corrosion depth on the surface of the sample corroded at at 800 m after the removal of correction scales.

Figure 11. Frequency density distribution map of the corrosion depths of samples corroded at different F_{rel} density distribution map of the corresponding F_{rel} of samples corresponding at samples corresponding at F_{rel} at F_{rel} at F_{rel} well depths.

Table 5 shows the fitting parameters for the curve shown in Figure 11. The parameters w[ere](#page-9-2) a correlation coefficient of 0.9886. Corrosion depths followed Gaussian distribution. The physical interpretation of corrosion depth distribution revealed that $y_0 = 0$ and $A = 1$ in Equation (6). obtained by using the Gaussian model (Equation (6)) multicurve global mode in origin 9.0 and interpretation of corrosion depth distribution revealed that y0 = 0 and A = 1 in Equation (6).

$$
y = y_0 + \frac{A}{w\sqrt{\pi/2}}e^{-2\frac{(x-x_c)^2}{w^2}}
$$
 (6)

unction; y_0 is the offset, $y_0 = 0$; A is the area, where $A = 1$; x is corrosion depth (μ m); x_c is the expected value (μ m); w is twice the standard deviation; x_c determines the location of the distribution curve; and w determines the amplitude of the distribution curve. the location of the distribution curve; and w determines the amplitude of the distribution curve. where *y* is the probability density function; y_0 is the offset, $y_0 = 0$; *A* is the area, where *A* = 1; *x* is

Table 5. Parameters for the Gaussian model fitting of corrosion depth distribution.

Well Depth/m		240	580	800	1050	1500
x_c / μ m	40.250	27.680	71.189	67.189	47.268	52.028
w	23.923	22.363	45.154	42.617	28.408	45.292

Table [6](#page-10-0) shows the results of variance analysis through the Gaussian model used to fit the curve shown in Figure [11.](#page-9-1) The Prob > F value of less than 0.01 indicates that the frequency density distribution curve in Figure [11](#page-9-1) shows Gaussian distribution. In Gaussian distribution, x_c indicates the average value of the random variables and represents the average corrosion depth. The corrosion of the sample surface may have reduced the datum plane during image acquisition. Moreover, the trend followed by x_c with the change in well depth differed from that followed by the average corrosion rate, especially when the average corrosion rate was high. By adding/subtracting a constant value, the frequency density curve for new random variables generated from random variables with Gaussian distribution was transformed into the translated frequency density curve for former variables in the *x*-direction without changing the shape of the frequency density curve. Therefore, the absence or presence of the datum plane during image acquisition will not affect the w of the fitting results for Gaussian distribution. The trend followed by w with the change in well depth was the same as that followed by the maximum corrosion depth. This similarity indicates the existence of a strong linear correlation between w and corrosion type, as illustrated in Figure [12.](#page-10-1) Moreover, w was positively related to the span of the frequency density curve. A small w value represents a density distribution curve with a narrow range and limited corrosion depth distribution. These characteristics indicate uniform corrosion. A high *w* value represents a density distribution curve with a broad range and corrosion depth with broad distribution. These characteristics indicate local corrosion.

Statistics		DF	Sum of Squares	Mean Square	F Value	Prob > F
	Regression	12	8.11×10^{-3}	6.76×10^{-4}	719.55	
Global	Residual	54	5.07×10^{-5}	9.39×10^{-7}		
	Uncorrected Total	66	8.16×10^{-3}			
	Corrected Total	60	5.11×10^{-3}			

Table 6. Results for variance analysis through Gaussian model fitting. Table 6. Results for variance analysis through Gaussian model fitting.

Figure 12. Relationship between the maximum corrosion depth and *w.*

4. Conclusions

Based on the observed corrosion behavior of J55 carbon steel in $CO₂/30$ % crude oil/brine mixtures under the simulated conditions of different well depths (0–1500 m), we conclude the following:

- (1) The average corrosion rate of J55 carbon steel initially increased and then decreased in the $CO₂/crude$ oil/brine environment as partial $CO₂$ pressure increased. Corrosion type shifted from uniform to local corrosion;
- (2) The main corrosion products on the surfaces of J55 carbon steel were FeCO₃ and CaCO₃;

(3) The distribution of corrosion depth obeyed Gaussian distribution, and *w* was positively correlated with the maximum corrosion depth.

Author Contributions: H.B., Y.W. and N.Z. conceived and designed the experiments; P.R. and H.B. were involved in carrying out the corrosion test, SEM and XRD experiments, analyzing the experimental data; H.B. and Y.M. analyzed the data and wrote the paper.

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