Ecofriendly Ultrasonic Rust Removal: An Empirical Optimization Based on Response Surface Methodology

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Abstract: This study shows that the hard-to-remove rust layer on the guide sleeve surface of a used cylinder can be removed using a specially developed, environmentally friendly formula for cleaning rust. Furthermore, we studied the rust removal technology that is based on ultrasonic cavitation and chemical etching. The surface morphology and structural components of the rust layer were observed using an electron microscope and an X-ray powder diffractometer. These tools were used to explore the mechanism of combined rust removal. Using response surface methodology (RSM) and central composite design (CCD), with the rust removal rate as our index of evaluation, data were analyzed to establish a response surface model that can determine the effect of cleaning temperature and ultrasonic power interaction on the rate of rust removal. Results showed that the main components of the rust layer on a 45 steel guide sleeve were α-FeOOH, γ-FeOOH, and Fe3O4. The rust was unevenly distributed with a loose structure, which was easily corroded by chemical reagents and peeled off under ultrasonic cavitation. With the increase in the cleaning temperature, the chemical reaction effect was intensified, and the cleaning ability was enhanced. With the increase in ultrasonic power, the cavitation effect was aggravated, the ultrasonic agitation was enhanced, and the rust removal rate was improved. According to response surface analysis and the application scope of the rust remover, we determined that the optimal cleaning temperature is 55 °C, and that the optimal ultrasonic power is 2880 W. The descaling rate under these parameters is 0.15 g·min⁻¹·m⁻².

Keywords: construction machinery; ultrasonic rust removal; response surface methodology; process parameters optimization; derusting rate

1. Introduction

There are a lot of pollutants such as rust on the surface of waste construction machinery parts. It is necessary to remove the pollutants on the surface of these parts to ensure their quality when they are inspected, repaired, and assembled [1,2]. At present, single- or multiple-combined processes, such as high-pressure water jet cleaning, steam cleaning, shot blasting, high-temperature roasting, chemical cleaning, and manual polishing, are usually used for treating pollutants [3–6]. However, steam cleaning and high-temperature roasting have high costs, while shot blasting easily damages the substrate. Manual polishing is low in efficiency and high in labor costs, and the cleaning effect is not obvious. We have independently developed a derusting test machine with an ultrasonic generator and a heating device to solve the problem of cleaning rust layers on surfaces. A special, environmentally friendly formula, with citric acid as the main body, has been developed and a process of ultrasonic rust removal has been used to remove the rust layer on the surface of the sample parts. Under the action of the sound field, the cavitation bubbles in the cleaning tank rapidly expand and contract, resulting in local high temperature and high...
pressure. This has the effect of destroying the rust layer on the sample’s surface. At the same time, the strong ultrasonic oscillation continuously agitates the cleaning liquid [7,8], accelerates the chemical interaction between the rust remover and the rust layer [9,10], and continuously dissolves and penetrates the dirt. At present, there have been many achievements in the field of ultrasonic cleaning and environmentally friendly rust removal. Zhang Baocai et al. [11] used ultrasonic compounding of molten salt cleaning technology to remove thick paint on the surface of remanufactured end caps. They combined the technology of chemical paint removal and ultrasonic cavitation, and explored the impact of cleaning temperature and ultrasonic power on the composite cleaning cycle. Wang Jian et al. [12] used the potential tracking method to study the dynamic changes in the ultrasonic pickling process of steel and the removal mechanism of the oxide layer. They found that the introduction of ultrasonic waves in the pickling process produced ultrasonic cavitation, which accelerated the reaction and greatly improved the rust removal rate. Lin Jinzhu [13] analyzed the physical and chemical properties of citric acid and the mechanism of rust removal. He pointed out the necessity and importance of using environmentally friendly acid to remove rust, laying a foundation for the establishment of a rust removal program.

A large number of single-factor tests have proved that cleaning fluid temperature and ultrasonic power have a direct effect on the rust removal rate. However, the joint effect of the two factors on the rust removal process is rarely studied. Response surface methodology (RSM) can be used to study the effects of one or more factors by facilitating the design of a reasonable test scheme, while the optimal conditions or results in the experimental design are found by analyzing the response surface or contours [14,15]. Therefore, this experiment intends to explore the combined effect of temperature and ultrasonic power through the response surface method. Response surface methodology mainly includes central composite design (CCD) and Box–Behnken experimental design (BBD). Of these two, the most widely used is central composite design [16,17]. In the CCD test design, the test points are composed of cube points, center points, and axial points [18,19], which are sequential, efficient, and flexible [20,21]. There are many practical applications of response surface methodologies in process parameter optimization. Yan Dongping et al. [22] used the central composite design to study the effect of process parameters in the milling experiment on the cutting force of the titanium alloy TC21. Yuan Julong et al. [23] optimized the polishing process of YG8 cemented carbide inserts via a response surface methodology so as to quickly determine the best process parameters of YG8 rake face polishing. Wang Qun et al. [24] used a response surface methodology to explore the effects of potassium ferrate dosage in flocculant and water pH on the water turbidity and UV254 removal rate, and optimized process parameters by establishing a secondary response model. Therefore, not only can the RSM establish a continuous mathematical model, but it can also show the interaction between factors, which is often used in process parameter optimization.

In this paper, the rust removal rate test was designed by combining RSM and CCD. A regression equation and a response surface model were established to study the effect of cleaning temperature and ultrasonic power on the rust removal rate. The optimal parameters of the rust removal process were found, and the cleaning technology was optimized, which provided theoretical support for the application of ultrasonic rust derusting technology in the remanufacturing cleaning field.

2. Experimental Procedure

2.1. Test Samples

A hydraulic cylinder guide sleeve with rust on its surface was used as the test sample. The guide sleeve was made of 45 steel and cut into 35 mm × 25 mm × 10 mm blocks for physical and chemical analysis of the rust layer. The corrosion morphology and the cross-section of the rust layer were observed using a Fei Inspection S50 scanning electron microscope (Thermo Fisher Scientific, Waltham, MA, USA), and the elements in the corrosion layer were analyzed using an Oxford X-act spectrometer. The structure and
composition of iron oxide in the rust layer were analyzed using a D8 ADVANCE X-ray
diffraactometer (Bruker, Karlsruhe, Germany). X-ray diffraction (XRD) measurements were
carried out using Cu targets and Kα radiation at 40 kV. The rust layer on the surface of the
sample was scraped off, and the scraped sample was fully ground in an agate mortar. This
was followed by sample preparation, and then test and result analysis.

2.2. Test Design

The sample for the process parameter optimization test was a uniformly rusted iron
sheet with a size of 40 mm × 40 mm × 1 mm. The derusting formula has a main body of
30‰ citric acid, with 6–10‰ sodium dodecylbenzene sulfonate, 6–10‰ JFC (fatty alcohol-
polyoxyethylene ether), and 6–10‰ benzotriazole added to perform a 45 s rust removal test.
At room temperature, the formula can clean 80% of the rust layer in 45 s. If the cleaning time
is extended, the derusting end point cannot be accurately determined, and the rust removal
effect of each group cannot be compared. The sample was weighed using FA2004 precision
electronic (Yoke Instrument, Shanghai, China) balance with a measurement accuracy of
0.1 mg. We used a self-made derusting test machine (Xuzhou Construction Machinery
Group, Xuzhou, China) with an ultrasonic generator and a heating device. The ultrasonic
power and cleaning temperature can be adjusted.

The lowest test temperature was the local annual average temperature. The maximum
test temperature was related to the material of the derusting test machine. The tank of
the machine is made of corrosion-resistant polypropylene material, and the temperature
resistance of its bonding part is about 80 °C. Since the machine is oriented to industrial
applications and has a high frequency of use, the reliability and safety of the machine are
particularly important. We chose 60 °C as the maximum heating parameter. The lowest
value of power was 0, which is the value where ultrasound was not applied. The maximum
power was the maximum value of the ultrasonic generator. Thus, the power regulation
range was 0–2880 W, and the cleaning temperature range was 20–60 °C.

3. Results and Discussion

3.1. Response Surface Methodology

Figure 1 shows the scanning electron microscope (SEM) spectra of the rust layer
surface and the cross-section, respectively. As shown in Figure 1a, the rust layer on the
surface of the sample is uneven, with a maximum cross-section thickness of 66.7 µm and a
minimum of 16.7 µm. There are a large number of irregular particles randomly distributed
on the surface of the sample. The size of the particles is uneven, and they are connected
to form a layer structure on the surface of the sample. The main components of 45 steel
are Fe, C, Si, and Mn. The electron diffraction spectroscopy (EDS) results indicate that
the elements of the rust layer are in accordance with 45 steel, with the exception of the O
content. This shows that the main component of the rust layer is an iron oxide compound.

Iron oxides mainly include (α-β-γ-δ-)Fe₂O₃, (α-β-γ-δ-)FeOOH, and Fe₃O₄, which have
different valence, crystal form, and structure. Therefore, in order to determine the compo-
nents of the rust layer, XRD was performed on the rust layer of the sample. The results
show that the main components of the rust products are α-FeOOH, γ-FeOOH, and Fe₂O₄.

According to the SEM and XRD data, the main components of the rust layer are
α-FeOOH, γ-FeOOH, and Fe₂O₄, and the rusted surface is loose and unevenly distributed.
Therefore, the pickling agent can be continuously introduced through the act of ultrasonic
pickling so that the rust layer can be peeled off from the sample surface. The mechanism of
ultrasonic pickling and rust removal is shown in Figure 2.
ultrasonic cavitation, the cavitation bubbles in the liquid shrink or expand rapidly, stable complexes, which sheds rust from the surface of the sample. Because of the substrate is destroyed by the strong oscillation from the ultrasonic waves. Because of the ultrasonic cavitation, the cavitation bubbles in the liquid shrink or expand rapidly, not only etches the rust on the surface of the sample but also chemically reacts with Fe^{2+} and Fe^{3+} in the rust layer. Moreover, organic acids will chelate with iron oxides to form stable complexes, which sheds rust from the surface of the sample.

On the basis of chemical etching, the bond between the surface oxide layer and the substrate is destroyed by the strong oscillation from the ultrasonic waves. Because of the ultrasonic cavitation, the cavitation bubbles in the liquid shrink or expand rapidly, resulting in a huge pressure to peel off the loose rust layer on the surface of the sample. In addition, the reacted pickling solution near the rust layer leaves the sample surface through ultrasonic stirring, and the unreacted pickling solution is replenished in time to ensure the efficient and continuous derusting reaction.

3.2. Central Composite Design

3.2.1. Central Compound Test Design

In this rust-cleaning process parameter optimization experiment, the central composite design is used to optimize the cleaning temperature (X) and ultrasonic power (Y). Taking the rust removal rate as the response value (Z), a two-level full factor test is established.
Four cubic points, five central points, and four axis points are selected to generate 13 sets of experiments. The factors and levels are shown in Table 1.

| Table 1. Test factor level table. |
|----------------------------------|
| **Factor** | **Code** | **Variable Level** |
| Cleaning temperature (°C) | X | −1 | −0.7 | 0 | 0.7 | 1 |
| Ultrasonic power (W) | Y | 0 | 432 | 1440 | 2448 | 2880 |

3.2.2. Rust Removal Test

1. According to the generated parameters, the cleaning temperature and ultrasonic power are adjusted. There are 13 groups of test parameters. The test factors and levels are shown in Table 2.

2. According to the order of each experimental group, the rusted iron sheets are numbered and cleaned with absolute alcohol. After drying, the rusted iron sheets are weighed and marked as \( m_0 \). The weighed iron sheets are then placed into the ultrasonic rust removal tank and cleaned at the specified temperature and ultrasonic frequency for 45 s. The descaling sheets are cleaned with anhydrous ethanol and weighed with electronic balance, which is marked as \( m_1 \).

3. According to Formula (1), the rust removal rate of a rusted iron sheet under various process parameters is calculated.

| Table 2. Center composite test design and the rust removal test results. |
|---------------------------|----------------|-----------------|-----------------|-----------------|
| **Number** | **X** | **Y** | **Cleaning Temperature/°C** | **Ultrasonic Power/W** | **Rust Removal Rate/(g min\(^{-1}\) m\(^{-2}\))** |
| 1 | 0 | 0 | 40 | 1440 | 0.087 |
| 2 | 0 | 1 | 40 | 2880 | 0.119 |
| 3 | 1 | 0 | 60 | 1440 | 0.158 |
| 4 | 0.7 | −0.7 | 54 | 432 | 0.095 |
| 5 | 0 | 0 | 40 | 1440 | 0.083 |
| 6 | 0 | 0 | 40 | 1440 | 0.098 |
| 7 | −0.7 | −0.7 | 26 | 432 | 0.044 |
| 8 | 0 | 0 | 40 | 1440 | 0.105 |
| 9 | 0 | −1 | 40 | 0 | 0.038 |
| 10 | −0.7 | 0.7 | 26 | 2448 | 0.090 |
| 11 | 0 | 0 | 40 | 1440 | 0.089 |
| 12 | 0.7 | 0.7 | 54 | 2448 | 0.153 |
| 13 | −1 | 0 | 20 | 1440 | 0.069 |

\[
V = \frac{m_0 - m_1}{S \times t} \quad (1)
\]

In the formula, \( v \) represents the rust removal rate (g min\(^{-1}\) m\(^{-2}\)); \( m_0 \) and \( m_1 \) represent the mass of the rusted sample before and after 45 s rust removal (g), respectively; \( S \) represents the sample area (m\(^2\)); and \( t \) represents the rust removal time (min).

In this experiment, the rust removal rate is used to describe the cleaning effect. The higher the rust removal rate within 45 s, the better the rust removal effect.

The results of the rust removal test and the center composite test design are shown in Table 2.

3.2.3. Regression Analysis

According to the results of central composite test, a quadratic regression model is established to describe the rust removal rate under different cleaning temperature and ul-
trasonic power conditions. The corresponding cleaning temperature and ultrasonic power are coded and analyzed. Table 3 shows the estimated regression coefficients, with coded units for analysis. X represents cleaning temperature, and Y represents ultrasonic power.

Table 3. Estimated regression coefficient of rust removal rate under various cleaning parameters.

| Term     | Coefficient | Standard Error of Coefficient | T-Value (abs.) | p-Value (abs.) |
|----------|-------------|-------------------------------|----------------|---------------|
| Constant | 0.0923      | 0.0036                        | 25.330         | 0.000         |
| X        | 0.0300      | 0.0029                        | 10.422         | 0.000         |
| Y        | 0.0223      | 0.0029                        | 7.739          | 0.000         |
| X* X     | 0.0098      | 0.0031                        | 3.179          | 0.016         |
| Y* Y     | -0.0048     | 0.0031                        | -1.540         | 0.167         |
| X* Y     | 0.0031      | 0.0041                        | 0.767          | 0.468         |

Through analysis, it was found that the p-values corresponding to the main effect of X and Y are (X = 0.000) and (Y = 0.000), which are less than the 0.05 level of significance. The original hypothesis is rejected and the regression is significant. The p-value of X* X is (X* X = 0.016), which is less than 0.05, and the impact is significant. The p-values of Y* Y and X* Y are (Y* Y = 0.167) and (X* Y = 0.468), greater than 0.05. The impact is not significant, so it should be removed from the quadratic regression model. Y* Y and X* Y were removed, and the adjusted coefficients were redistributed, as shown in Table 4.

Table 4. Estimated regression coefficient of rust removal rate under various cleaning process parameters of corroded iron sheet (removal of insignificant items).

| Term     | Coefficient | Standard Error of Coefficient | T-Value (abs.) | p-Value (abs.) |
|----------|-------------|-------------------------------|----------------|---------------|
| Constant | 0.0890      | 0.0031                        | 28.745         | 0.000         |
| X        | 0.0300      | 0.0030                        | 9.906          | 0.000         |
| Y        | 0.0223      | 0.0030                        | 7.356          | 0.000         |
| X* X     | 0.0104      | 0.0032                        | 3.240          | 0.010         |

Table 4 shows the results of the model analysis after removing the insignificant items. The p-values of each group are all less than 0.05, and the model is generally valid.

The analysis of variance is shown in Table 5. The two determination coefficients are (R-Sq = 94.76%) and (R-Sq (adjusted) = 93.01%). The difference between them is small and close to 1, which means that the regression is high. The cleaning cycle model is applicable.

Table 5. Variance analysis of rust removal rate under various cleaning process parameters (R-Sq = 94.76%, R-Sq (forecast) = 89.30%, R-Sq (adjustment) = 93.01%).

| Source    | Freedom | Seq SS  | Adj SS   | Adj MS   | F-Value (abs.) | p-Value (abs.) |
|-----------|---------|---------|----------|----------|---------------|---------------|
| Regression| 3       | 0.011967| 0.011967 | 0.003989 | 54.25         | 0.000         |
| Linear    | 2       | 0.011195| 0.011195 | 0.005597 | 76.12         | 0.000         |
| X         | 1       | 0.007216| 0.007216 | 0.007216 | 98.14         | 0.000         |
| Y         | 1       | 0.003979| 0.003979 | 0.003979 | 54.11         | 0.010         |
| Square    | 1       | 0.000772| 0.000772 | 0.000772 | 10.50         | 0.000         |
| X* X      | 1       | 0.000772| 0.000772 | 0.000772 | 10.50         | 0.000         |
| Error     | 9       | 0.000662| 0.000662 | 0.000074 | 0.91          | 0.553         |
| Pure error| 4       | 0.000330| 0.000330 | 0.000078 | 0.91          | 0.553         |
| Total     | 12      | 0.012629|          |          |               |               |

Table 6 shows the estimated regression coefficient of the rust removal rate under various cleaning process parameters of a rusted iron sheet (using uncoded unit data).
Table 6. Regression coefficient of the rust removal rate under various cleaning process parameters of a rusted iron sheet (using uncoded unit data).

| Term     | Coefficient |
|----------|-------------|
| Constant | 0.0560859   |
| X        | -0.00205385 |
| Y        | 0.0000219022 |
| X*X      | 0.0000522192 |

According to the calculation results after coefficient correction, the modified regression equation can be obtained. The influencing factors in the equation include the first-order effect and second-order effect of cleaning temperature and the second-order effect of ultrasonic power.

3.2.4. Residual Analysis

Residual refers to the difference between the actual observed value and the regression estimated value. Residual analysis is used to analyze the reliability, periodicity, or other disturbances of the data through the information provided by the residual and further characterizes the adaptability of the model equation. The main purpose of residual diagnosis is to diagnose whether the model fits well with the data based on the status of the residuals. The residual diagram of rust removal rate is shown in Figure 3.

![Residual Diagram](image)

Figure 3. Residual diagram of rust removal rate under various cleaning parameters.

As is shown in Figure 3, the residuals of different process parameters conform to normal distribution from the normal probability diagram of residuals. According to the scatter plot with the predicted value of the response variable fitting to the horizontal axis of the residual, the residuals maintain equal variances and have no obvious regularity. There is no ‘funnel shape’ or ‘trumpet shape’. According to the residuals for the scatter plot with the order of observations as the horizontal axis, each point randomly fluctuates up and down the horizontal axis. In summary, the residuals under different process parameters are not abnormal.
3.2.5. Establishment of Regression Equation

After model adjustment, variance analysis, and residual diagnosis, the regression equation of rust removal rate under various cleaning parameters of the corroded iron sheet is obtained, as shown in Equation (2).

\[
RT = 0.0560859 - 0.00205385x + 0.0000219022y + 0.0000522192x^2 + \epsilon \tag{2}
\]

Among them, \(RT\) represents the rust removal rate, \(x\) represents the cleaning temperature, \(y\) represents ultrasonic power, and \(\epsilon\) represents error.

3.2.6. Response Surface

The response surface of rust removal rate is shown in Figure 4.

\[\text{Figure 4. Evolution of rust removal rate with respect to cleaning temperature and ultrasonic power under various cleaning parameters.}\]

The light green area in the upper right corner of Figure 4 shows a good rust removal rate, which is greater than 0.175 g·min\(^{-1}\)·m\(^{-2}\). It can be seen from Figure 5 that the higher the cleaning temperature, the higher the rust removal rate, and the greater the ultrasonic power, the higher the rust removal rate, which is consistent with the phenomenon observed in the test. This behavior can be explained by considering the aqueous ultrasonic bath type, i.e., the acidic solution. The acid reacts with the rust layer in the rust removal tank, and the hydrogen bubbles thus generated have a peeling effect on the rust layer \([25,26]\). When the temperature increases, the chemical reaction rate increases, resulting in a faster rust removal rate. The vibration of the ultrasonic vibrator in the rust removal tank generates ultrasonic waves, and the ultrasonic cavitation effect appears in the liquid. When the sound energy reaches the critical point, the cavitation nucleus breaks down and produces a huge impact force, which causes the dirt on the liquid surface to decompose and peel off \([27]\). At the same time, due to the effect of micro-acoustic flow and local high-pressure impact, the ultrasonic high-speed stirring rust removal solution accelerates the chemical reaction on the surface of the sample, which accelerates the dissolution rate of the dirt \([28]\). When the ultrasonic power in the sound field increases, the cavitation effect and ultrasonic oscillation effect are intensified. The reaction between the rust remover and the rust layer is accelerated, and the rust removal rate increases.

When the cleaning temperature is at a low level, the response surface increases greatly with the increase in ultrasonic power. When the cleaning temperature is high, the increase in the response surface is small with the increase in ultrasonic power. Therefore, according to the regression equation, when the cleaning temperature is 20 °C, the rust removal efficiency is 0.036 g·min\(^{-1}\)·m\(^{-2}\) without ultrasonic power. When ultrasonic power is applied, the rust removal efficiency is 0.099 g·min\(^{-1}\)·m\(^{-2}\), which is 2.76 times greater than the former. When the ultrasonic power is low, the increase in the response surface is larger. When the ultrasonic power is high, the increase in the response surface is small with the
increase in cleaning temperature. Therefore, according to the regression equation, when the ultrasonic power is 0, the rust removal efficiency is 0.036 g·min$^{-1}$·m$^{-2}$ without heating. After heating, the rust removal efficiency is 0.12 g·min$^{-1}$·m$^{-2}$, which is 3.37 times greater than the former. Thus, in the experimental range, the influence of temperature on the rust removal rate of 45 steel is greater than that of ultrasonic power.

![Figure 5. Optimization of rust removal rate under various cleaning parameters.](image)

3.2.7. Response Optimization

The response variable optimizer is used to obtain the minimization optimization result of the cleaning cycle of the rusty iron sheet, as shown in Figure 5.

As shown in Figure 5, the optimal value of the rust removal rate is reached when the cleaning temperature is 60 °C and the ultrasonic power is 2880 W. At this point, the rust removal rate is 0.1827 g·min$^{-1}$·m$^{-2}$, and the desirability (d) is 1.0000.

3.2.8. Result Verification

With the increase in temperature, the speed of chemical reaction and rust removal increases. However, in practical applications, the increase in temperature will lead to the corrosion of the substrate material by the acidic rust remover, which causes hydrogen embrittlement. Results show that the pickling temperature is generally controlled within 60–70 °C. Therefore, considering the rust removal rate and the application scope of rust remover, we can reduce the temperature appropriately and choose 55 °C as the optimal rust-cleaning temperature.

The cleaning temperature should therefore be set to 55 °C and the ultrasonic power to 2880 W. The experiment was designed to ultrasonically clean the sample cylinder guide sleeve for 45 s. Figure 6 shows the micromorphology of the sample before and after cleaning. The results show that most of the rust on the sample surface was removed after cleaning. The rust removal rate at this time is calculated to be 0.15 g·min$^{-1}$·m$^{-2}$.
sleeve for 45 s. Figure 6 shows the micromorphology of the sample before and after cleaning. The results show that most of the rust on the sample surface was removed after cleaning. The rust removal rate at this time is calculated to be 0.15 g·min$^{-1}$·m$^{-2}$.

Figure 6. SEM surface morphology of the cylinder guide sleeve: (a) before rust cleaning; (b) after rust cleaning.

4. Conclusions

1. This paper takes the rust of a cylinder guide sleeve as an example of how to optimize rust removal efficiency and use the environmentally friendly citric acid as an alternative to traditional cleaning chemicals for rust removal. Under the action of H$^+$ and ultrasonic cavitation impact, the rust layer reacts and peels off.

2. The regression equation and response surface model of rust removal rate were obtained by using a central composite test method. The higher the cleaning temperature and the ultrasonic power, the higher the rust removal rate. Considering the rust removal rate and the application scope of rust remover, we chose 55 °C as the optimal rust-cleaning temperature.

3. The optimal process parameters of ultrasonic rust removal have been determined. The cleaning temperature is 55 °C, the ultrasonic power is 2880 W, and the descaling rate under the optimal parameters is 0.15 g·min$^{-1}$·m$^{-2}$.

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