

Process optimization and economic calculation of photocatalytic degradation of methylene blue over sulfur-doped g-C₃N₄

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Abstract: Sulfur doping has been considered an effective strategy to enhance the photocatalytic activity of graphitic carbon nitride (g-C₃N₄). However, there are no reports on the process optimization and economic calculation to determine the optimal photocatalytic parameters and compare the operating costs, respectively. In this work, sulfur-doped g-C₃N₄ (S/g-C₃N₄) was prepared *in situ* by a one-pot pyrolysis approach using thiourea as the precursor and sulfur source. S/g-C₃N₄ was characterized by various instrumental techniques. Response surface methodology (RSM) was employed to determine the optimal photocatalytic conditions for the degradation of Methylene blue (MB) over S/g-C₃N₄. A central composite design (CCD) with four factors at five levels was established, comprising 30 experimental runs. The optimal conditions were determined as an initial MB concentration of 27.6 mg L⁻¹, a catalyst dosage of 4.1 g L⁻¹, a solution pH of 8.7 and a reaction time of 127 min. Under these conditions, the predicted degradation efficiency could reach 99.94 %, while the experimental value was 99.5 %. Cost accounting indicated that the optimization substantially reduced direct costs, greatly improved equipment utilization, and enhanced operational flexibility. This work will lay key foundation for cost control and efficiency improvement in the scale-up application of g-C₃N₄-based photocatalysts.

Keywords: graphitic carbon nitride; photocatalyst; response surface methodology; cost accounting.

INTRODUCTION

Resource shortages and environmental pollution are closely related to our quality of life and health conditions, and have drawn considerable attention from governments around the world. Among the two, the issues caused by environmental pollution are more significant.¹ Among various kinds of environmental pollution, organic dye wastewater has attracted the most attention because of its high

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34 concentration, diverse composition, difficult decolorization, widespread pollution
35 and complex structure. At present, there are many technologies to treat organic dye
36 wastewater, including adsorption, membrane separation, biodegradation, extraction,
37 ion, oxidation and so on.² In recent years, the development of advanced oxidation
38 technologies represented by photocatalytic oxidation has attracted considerable
39 attention. The use of photocatalytic materials to purify water has become a high-
40 tech environmental purification technology of global concern.³

41 In recent decades, a variety of novel visible-light-response photocatalytic materials
42 have been developed, such as bismuth oxyhalides, metal-organic frameworks,
43 ZnIn_2S_4 , CuFeS_2 , WO_3 , CuWO_4 , Cu_2SnS_3 and others. These materials have high
44 visible-light catalytic stability and activity.⁴⁻⁷ However, the precious metals in the
45 components increase the preparation and production costs of photocatalytic materials.
46 At the same time, they have certain toxicity and the potential to pollute the
47 water environment. Therefore, these materials are not suitable for large-scale practical
48 applications. Graphitic carbon nitride ($\text{g-C}_3\text{N}_4$) is a new type of non-metallic
49 semiconductor with a band gap of about 2.7 eV, which can achieve visible light
50 response and make full use of solar energy. Moreover, $\text{g-C}_3\text{N}_4$ has a stable structure,
51 controllable performance, a non-toxic and harmless nature, acid and alkali
52 resistance, and light corrosion resistance. It has become a hotspot in the research
53 field of photocatalytic materials.^{8,9} However, the shortcomings of $\text{g-C}_3\text{N}_4$, such as
54 fast electron-hole recombination and a small specific surface area, limit its practical
55 application. Therefore, heterojunction construction, doping modification, microstructure
56 regulation and other methods have been employed to improve the photocatalytic activity
57 of $\text{g-C}_3\text{N}_4$.^{10,11} Among them, doping with non-metallic elements could effectively change
58 the electronic band structure of $\text{g-C}_3\text{N}_4$, thereby improving its photocatalytic activity.¹²
59 Xu and coworkers successfully prepared $\text{g-C}_3\text{N}_4$ with nitrogen defects and sulfur doping
60 via a cold plasma technique. S-doping could offer more active sites for the photocatalytic
61 process and enhance the stability of N defects. The visible light photocatalytic activity
62 of S-doped $\text{g-C}_3\text{N}_4$ was 11.25 times higher than that of pure $\text{g-C}_3\text{N}_4$.¹³ Ma *et al.* used
63 the dielectric barrier discharge (DBD) plasma to obtain an S-doped $\text{g-C}_3\text{N}_4$ photocatalyst
64 under an H_2S atmosphere. They found that DBD plasma treatment could make more
65 sulfur be doped into the $\text{g-C}_3\text{N}_4$ lattice than traditional roasting approach. The as-prepared
66 S-doped $\text{g-C}_3\text{N}_4$ exhibited much higher photocatalytic degradation efficiency of rhodamine
67 B (RhB).¹⁴ In addition, a porous S-doped $\text{g-C}_3\text{N}_4$ was successfully prepared via one-step
68 calcination method by Yang and coworkers. The as-prepared porous S-doped $\text{g-C}_3\text{N}_4$
69 displayed approximately 6.2 times higher peroxydisulfate activation capacity for the
70 degradation of RhB than pristine $\text{g-C}_3\text{N}_4$ under visible light irradiation.¹⁵ Song *et al.*
71 employed urea and thiourea as precursors to prepare S-doped $\text{g-C}_3\text{N}_4$ nanosheets
72 through physical steam activation. The as-prepared photocatalyst exhibited 2.1 and 5.52
73 fold higher

decomposition efficiency and reaction rate for the photodegradation of Methylene blue (MB) dye than bulk g-C₃N₄, respectively.¹⁶ Overall, S doping has been proven to be an efficient strategy to improve light absorption and photogenerated electron–hole separation, consequently boosting the photocatalytic activity of g-C₃N₄.¹⁷ However, there are few reports on the process optimization of photocatalytic degradation of organic pollutants over S-doped g-C₃N₄. Response surface methodology (RSM) based on a central composite design (CCD) is a widely used process parameter optimization method.^{18–20} This methodology includes experimental condition design and parameter optimization. The specific steps of this methodology can be divided into the following four aspects:^{21,22} 1) determination of the variables and experimental design, 2) evaluation of the parameters for mathematical model regression analysis, 3) prediction of the model response and 4) confirmation of the accuracy of the model. In addition, as far as we know, the economic calculation of photocatalytic degradation of organic pollutants over S-doped g-C₃N₄ before and after process optimization has not been investigated.

Therefore, in this work, S-doped g-C₃N₄ was prepared by a facile one-pot pyrolysis approach using thiourea as the precursor and sulfur source to realize *in situ* S doping and achieve high photocatalytic activity. Based on the characterization analyses and photocatalytic performance evaluation of S-doped g-C₃N₄, the RSM method was employed for the first time to optimize the process of MB photodegradation over S-doped g-C₃N₄ to determine the optimal operating parameters. Moreover, the economic calculations before and after optimization were performed for the first time. Therefore, this work will lay an engineering and cost foundation for the large-scale practical applications of g-C₃N₄-based photocatalysts.

EXPERIMENTAL

Preparation of S-doped g-C₃N₄

A facile one-pot pyrolysis approach was employed to prepare *in situ* S-doped g-C₃N₄ using thiourea as the precursor and sulfur source.²³ The specific procedures are as follows: 15 g of thiourea was weighed and put into a 100 mL crucible, and then heated from room temperature to 550 °C at a heating rate of 20 °C/min in a muffle furnace. After being kept for 4 h, the sample was cooled to room temperature in the furnace. Finally, it was ground into powder for subsequent use.

Characterization techniques

X-ray diffraction (XRD), Fourier transform infrared spectroscopy (FTIR), field emission scanning electron microscopy (FESEM), transmission electron microscopy (TEM) and UV–Vis diffuse reflectance spectroscopy (UV–Vis DRS) were used to characterize the as-prepared samples. The detailed information is provided in the supplementary file.

Photocatalytic experiments

The photocatalytic performance of S-doped g-C₃N₄ was evaluated using MB dye as the target pollutant. The experimental procedure is provided in the Supplementary material to this paper.

RSM optimization method

In this work, four parameters were selected as variables, namely the initial MB concentration (mg L^{-1}), catalyst dosage (g L^{-1}), solution pH and reaction time (min), which were labeled as X_1 , X_2 , X_3 and X_4 , respectively. At the same time, the degradation rate of MB dye was used as the output response (Y). The detailed descriptions of experimental design, mathematical modeling and response optimization are provided in the Supplementary material. In addition, the range and the level of independent variables are shown in Table S-I of the Supplementary material.

Economic calculation method

Generally, the standardized economic calculation for a wastewater treatment process included three main parts: raw material cost, operation cost and integrated cost.²⁴⁻²⁶ Their calculation methods are provided in the Supplementary material.

RESULTS AND DISCUSSION

Characterization of S-doped $\text{g-C}_3\text{N}_4$

Fig. 1A presents the XRD pattern of the S-doped $\text{g-C}_3\text{N}_4$ sample. The XRD pattern shows two prominent diffraction peaks located at 2θ 13 and 27.5° . The peak at 13° belongs to the in-plane ordering of triazine units and is indexed to the (100) crystal plane of $\text{g-C}_3\text{N}_4$, whereas the peak at 27.5° corresponds to the inter-layer stacking of aromatic layers and is assigned to the (002) crystal plane of $\text{g-C}_3\text{N}_4$.²⁷ The FT-IR spectrum of S-doped $\text{g-C}_3\text{N}_4$ is displayed in Fig. 1B. The absorption bands in the $1225\text{--}1640\text{ cm}^{-1}$ region are assigned to stretching vibrations of C–N and C=N bonds within the N-containing ring framework, together with exocyclic C–N stretching.²⁸ The broad band at 3135 cm^{-1} is associated with N–H stretching vibrations,²⁹ while the band at 796 cm^{-1} is attributed to the bending vibration of the C–N ring.³⁰

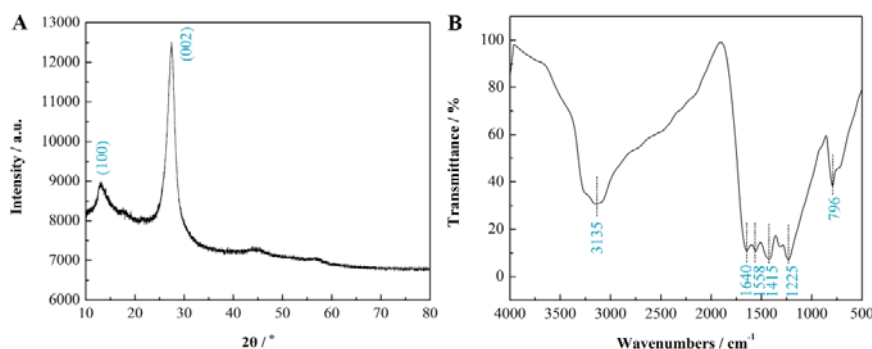


Fig. 1. A) XRD pattern and B) FTIR spectrum of S-doped $\text{g-C}_3\text{N}_4$.

Fig. 2A gives an FESEM image of S-doped $\text{g-C}_3\text{N}_4$, displaying a non-uniform particulate morphology in which the particles are irregularly agglomerated, consistent with the typical morphology of $\text{g-C}_3\text{N}_4$.³¹ Fig. 2B shows a TEM image of the

146 S-doped g-C₃N₄ sample. It can be seen that the as-prepared sample has a typical
 147 thin-layer structure with some wrinkles, similar to graphene. This thin-layer struc-
 148 ture will further increase the specific surface area of the S-doped g-C₃N₄ sample,
 149 thereby improving the adsorption capacity and photocatalytic activity of the sam-
 150 ple.³² From the energy-dispersive spectrum (EDS) of S-doped g-C₃N₄ (Fig. 2C),
 151 it can be clearly seen that the as-prepared sample contains three main elements C,
 152 N and S, which indicates that S has been successfully introduced into g-C₃N₄. In
 153 addition, no characteristic diffraction peak or absorption peak belonging to S was
 154 found in the XRD pattern and the FTIR spectrum, which implies that S might enter
 155 the crystal structure of g-C₃N₄ and replace C or N, rather than as impurities.^{13–16}

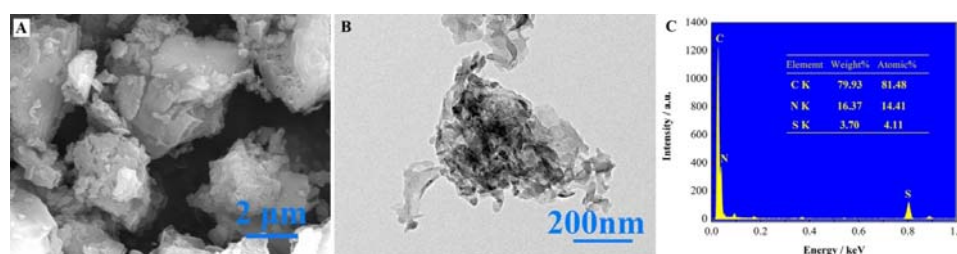


Fig. 2. A) FESEM, B) TEM and C) EDS images of S-doped g-C₃N₄.

158 Fig. 3A shows the UV–Vis DRS spectrum of S-doped g-C₃N₄, from which it
 159 can be observed that the S-doped g-C₃N₄ sample has strong visible-light absorpt-
 160 ion and an absorption edge is located at approximately 412 nm. According to the
 161 extrapolation of $(F(R)-h\nu)^{1/2}$ curve vs. photon energy $h\nu$ (Fig. 3B), the band gap
 162 (E_g) of S-doped g-C₃N₄ is estimated to be 2.63 eV, in agreement with literature
 163 reports.³³ Notably, the as-prepared sample exhibits an additional absorption fea-
 164 ture near 645 nm in the UV–Vis DRS spectrum, corresponding to an E_g of 1.81 eV.
 165 This might be associated with an S-induced impurity energy level in the band
 166 structure,³⁴ which can facilitate the enhancement of photocatalytic performance.

167 Process optimization of MB photodegradation over S-doped g-C₃N₄

168 To evaluate the photocatalytic performance of S-doped g-C₃N₄, MB with
 169 analytical purity was used as the target pollutant under the conditions of 10 mg L⁻¹
 170 MB, 5.0 g L⁻¹ S-doped g-C₃N₄, pH 7 and room temperature. As shown in Fig. 4A,
 171 after 60 min in the dark to reach adsorption equilibrium, the removal rate of MB is
 172 only 9.4 %, suggesting that the contribution of adsorption to MB removal can be
 173 neglected. However, after 120 min of photocatalytic irradiation, the corresponding
 174 degradation rate increased to 95.8 %. Accordingly, the as-prepared S-doped g-C₃N₄
 175 sample exhibits stronger photocatalytic performance toward MB degradation.
 176 Additionally, this photocatalytic process follows the pseudo-first-order kinetic
 177 model (Fig. 4B), with a reaction rate constant of 0.02296 min⁻¹. The UV–Vis
 178 absorption spectra of the MB solution during different photocatalytic periods are

presented in Fig. 4C, where it can be seen that the main absorption peak of the MB dye molecule at 650 nm disappeared after 120 min of photocatalytic reaction, indicating that the dye molecule has been decomposed. In addition, the removal percentage of total organic carbon (TOC) after 120 min of photocatalytic irradiation was measured as 78.4 %. Therefore, it can be supposed that the MB dye was decomposed into CO₂, H₂O, and other low-molecular-weight substances.

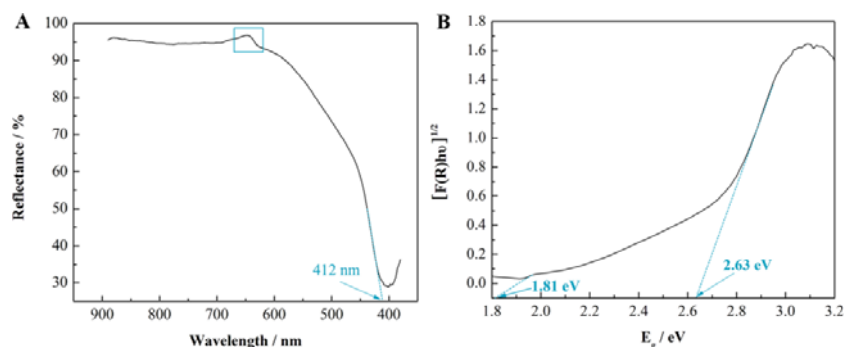


Fig. 3. A) UV-Vis DRS spectrum and B) $(F(R)-h\nu)^{1/2}$ versus $h\nu$ curve of S-doped g-C₃N₄.

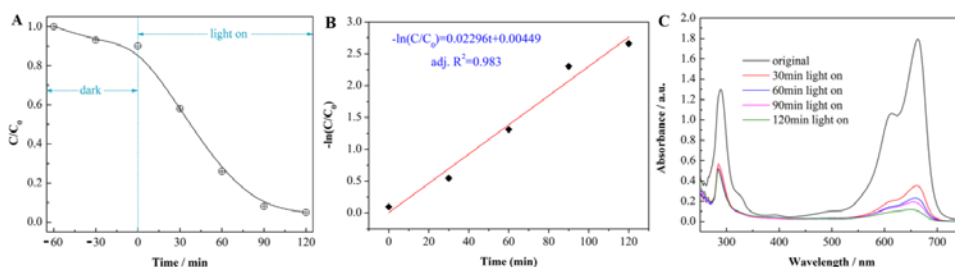


Fig. 4. A) Photocatalytic performance toward MB degradation over S-doped g-C₃N₄; B) fitting line of the pseudo-first-order kinetic model; C) UV-Vis absorption spectra of the MB solution during different photocatalytic periods.

Based on the above results, the effects of different experimental parameters on the degradation rate of MB dye were evaluated by RSM. According to the CCD, an experimental scheme with 4 factors and 5 levels was designed, and a total of 30 sets of experiments were carried out. The CCD experimental matrix is listed in Table S-II of the Supplementary material, along with the experimental and predicted values of the response Y . Among them, six groups of experiments are replicate experiments at the center point, which ensures that the relative error of the experiment is small. Their close response values mean higher experimental accuracy.³⁵

To predict the response value, the experimental values were fitted by a second-order polynomial regression model, and the results are shown as:

$$\begin{aligned}
Y = & 66.46 + 5.82X_1 - 13.35X_2 + 7.62X_3 + 12.70X_4 - 2.74X_1X_2 + \\
& + 1.69X_1X_3 + 1.70X_1X_4 + 0.79X_2X_3 + 0.21X_2X_4 - 1.17X_3X_4 - \\
& - 2.48X_1^2 - 3.90X_2^2 - 10.22X_3^2 - 2.61X_4^2
\end{aligned} \quad (1)$$

From the analysis of variance (ANOVA, Table S-III of the Supplementary material), it can be seen that the F value is 8.09, which is much higher than the standard value $F_{0.05} = 2.42$, which means that the second-order polynomial model is highly significant. In addition, as listed in Table S-II, the experimental and predicted values of the response Y are relatively close, with a correlation coefficient of $R^2 = 0.9588$. This is the main criterion for evaluating the correlation between the experimental and predicted values.³⁶

Moreover, the predicted response values are in good agreement with the experimental values (Fig. S-2 of the Supplementary material). Furthermore, the $Prob > F$ value is less than 0.0500, which means that the influence of the variable is significant, whereas the value greater than 0.1000, indicates that the influence of the variable is not significant.³⁷ Therefore, in this work, it can be seen from Table S-III that X_1 , X_2 , X_3 , X_4 , X_1X_3 , X_1X_4 , X_1^2 , X_3^2 and X_4^2 are the significant influencing variables during the MB photodegradation over S-doped g-C₃N₄. In addition, the statistical analysis of the experimental data indicates that the initial MB dye concentration, S-doped g-C₃N₄ dosage, solution pH and reaction time have significant effects on MB photodegradation.

The three-dimensional (3D) response surface plots of the interaction effect between two factors are shown in Fig. 5, including the interactions between initial MB concentration and S-doped g-C₃N₄ dosage, initial MB concentration and solution pH, initial MB concentration and reaction time, S-doped g-C₃N₄ dosage and solution pH, S-doped g-C₃N₄ dosage and reaction time and solution pH and reaction time. It can be seen that the degradation rate of MB increases with the decrease in the initial MB concentration and the increase in the S-doped g-C₃N₄ dosage (Fig. 5A); increases with the decrease in the initial MB concentration and the increase in the solution pH (Fig. 5B); increases with the decrease in the initial MB concentration and the increase of the reaction time (Fig. 5C); increases with the increase of the S-doped g-C₃N₄ dosage and the solution pH (Fig. 5D); increases with the increase in the S-doped g-C₃N₄ dosage and the reaction time (Fig. 5E); increases with the increase in the reaction time and the solution pH (Fig. 5F). According to the prediction model, the process parameters of MB photodegradation over the S-doped g-C₃N₄ photocatalyst are optimized as follows: the initial MB concentration is 27.6 mg L⁻¹, the S-doped g-C₃N₄ dosage is 4.1 g L⁻¹, the solution pH is 8.7, and the reaction time is 127 min. Under these optimal conditions, the predicted degradation rate of MB can reach 99.94 %. Under the identical conditions, the experimental value of the MB degradation rate is 99.5 %, showing excellent agreement with the model prediction.

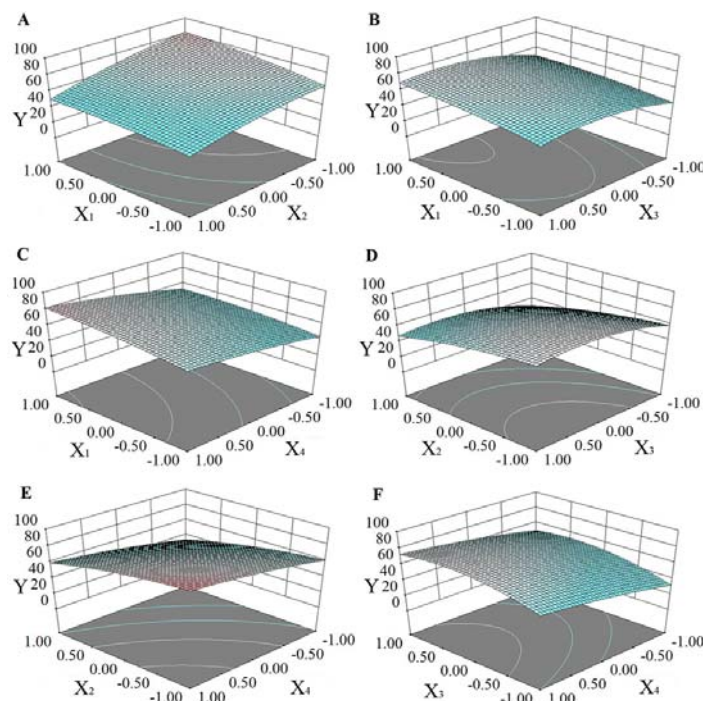


Fig. 5. 3D response surface of MB photodegradation (Y) over S-doped $g\text{-C}_3\text{N}_4$ with the interaction effect between: A) initial MB concentration (X_1) and S-doped $g\text{-C}_3\text{N}_4$ dosage (X_2), B) initial MB concentration (X_1) and solution pH (X_3), C) initial MB concentration (X_1) and reaction time (X_4), D) S-doped $g\text{-C}_3\text{N}_4$ dosage (X_2) and solution pH (X_3), E) S-doped $g\text{-C}_3\text{N}_4$ dosage (X_2) and reaction time (X_4) and F) solution pH (X_3) and reaction time (X_4).

Economic calculation of MB photodegradation over S-doped $g\text{-C}_3\text{N}_4$

1) *Comparative analysis of raw material cost.* Table S-IV of the Supplementary material provides the comparative analysis of raw material cost before and after optimization. Before optimization, the catalyst cost for treating 1 m^3 of MB wastewater was $5\text{ kg} \times 80\text{ CNY kg}^{-1} = 400\text{ CNY}$. After optimization, the corresponding cost for treating 1 m^3 of MB wastewater was $4.1\text{ kg} \times 80\text{ CNY kg}^{-1} = 328\text{ CNY}$. Therefore, the catalyst cost for treating 1 m^3 wastewater was reduced by 72 CNY, with a decrease of 17.8 %. Moreover, after optimization, the pH value of MB solution was adjusted from 7.0 to 8.7, so sodium hydroxide needed to be added to adjust the pH value. Based on the calculation of the acid-base balance, 0.12 kg of sodium hydroxide was required to adjust the pH from 7.0 to 8.7 for 1 m^3 wastewater, and the corresponding cost was $0.12\text{ kg} \times 2.5\text{ CNY kg}^{-1} = 0.3\text{ CNY}$. After optimization, although the acid-base adjustment cost was slightly increased, combined with the significant reduction in catalyst cost, the integrated raw material

cost still achieved a net saving of 71.7 CNY for treating 1 m³ wastewater, with a reduction of 17.7 %.

2) *Comparative analysis of operation cost.* Table S-V of the Supplementary material gives the comparative analysis of operation cost before and after optimization. The energy consumption of photocatalytic reaction mainly came from the xenon lamp (300 W) and the stirrer (50 W). The energy consumption was positively correlated with the reaction time. Before optimization, the energy consumption for treating 1 m³ wastewater was $(300\text{ W} + 50\text{ W}) \times 3\text{ h} = 1.05\text{ kWh}$ and the corresponding energy cost was $1.05\text{ kWh} \times 0.6\text{ CNY kWh}^{-1} = 0.63\text{ CNY}$. After optimization, the reaction time was 127 min (2.12 h). Therefore, the energy consumption for treating 1 m³ wastewater was $(300\text{ W} + 50\text{ W}) \times 2.12\text{ h} = 0.742\text{ kWh}$, and the corresponding energy cost was $0.742\text{ kWh} \times 0.6\text{ CNY kWh}^{-1} = 0.445\text{ CNY}$. Therefore, after optimization, the energy consumption cost for treating 1 m³ wastewater was reduced by 0.185 CNY, with a decrease of 29.4 %.

With regard to the labor cost, before optimization, assuming that the operation time of a single batch treatment of 1 m³ wastewater was about 3.5 h, the labor cost was $3.5\text{ h} \times 30\text{ CNY h}^{-1} = 105\text{ CNY}$. After optimization, the operation time was shortened to 2.5 h, and the labor cost was $2.5\text{ h} \times 30\text{ CNY h}^{-1} = 75\text{ CNY}$. Therefore, the labor cost for treating 1 m³ wastewater was directly reduced by 30 CNY, with a decrease of 28.6 %.

3) *Comparative analysis of integrated cost.* Based on the raw material and operation costs, the integrated cost for treating 1 m³ MB wastewater was calculated. Before optimization, the integrated cost for treating 1 m³ MB wastewater was $(400 + 0 + 0.63 + 105)\text{ CNY} = 505.63\text{ CNY}$. After optimization, the integrated cost for treating 1 m³ MB wastewater was $(328 + 0.3 + 0.445 + 75)\text{ CNY} = 403.745\text{ CNY}$. Therefore, after optimization, the integrated cost for treating 1 m³ MB wastewater was reduced by 101.885 CNY, with a decrease of 20.0 %.

CONCLUSION

In this work, S-doped g-C₃N₄ was successfully prepared by a facile one-step pyrolysis approach, using thiourea as the precursor and sulfur source. The as-prepared sample was characterized by XRD, FTIR, FESEM, TEM, EDS and UV-Vis DRS. The sample exhibited a wide visible-light absorption region, and its E_g was estimated to be 2.63 eV. Furthermore, an additional absorption feature near 645 nm was observed in the UV-Vis DRS spectrum, corresponding to an E_g of 1.81 eV. This might be associated with an S-induced impurity energy level in the band structure. The as-prepared sample exhibited stronger photocatalytic performance toward MB degradation. After 120 min of photocatalytic irradiation, the MB degradation rate reached 95.8 %. Based on RSM under CCD, the process optimization of MB photodegradation over the S-doped g-C₃N₄ photocatalyst was

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SUPPLEMENTARY MATERIAL TO
Process optimization and economic calculation of photocatalytic degradation of methylene blue over sulfur-doped g-C₃N₄

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EXPERIMENTAL

Characterization techniques

The crystal phase composition of the product was determined by X-ray diffraction (XRD), conducted on a Rigaku D/max-3B X-ray diffractometer (Japan). The functional groups of the sample were analyzed by Fourier transform infrared spectroscopy (FTIR), operated on a Nicolet Nexus infrared spectrometer (United States). The microscopic morphology of the sample was observed by field emission scanning electron microscopy (FESEM), realized on a FEISirion 200 scanning electron microscope (Netherlands). The microstructure of the sample was observed by transmission electron microscopy (TEM), obtained on a JEOL JEM-2010 transmission electron microscope (Netherlands). The optical absorption edge and band gap of S-doped g-C₃N₄ were analyzed by UV-vis diffuse reflectance spectroscopy (UV-vis DRS), recorded on an Ocean Optics USB4000 UV-visible diffuse reflectance spectrometer with integrating sphere, using a South African optical standard template was used as the reference. The content of total organic carbon (TOC) was measured by a XPERT-TOC/TNb TOC analyzer (Netherlands).

Photocatalytic experiments

The photocatalytic performance of S-doped g-C₃N₄ was evaluated under a 300W xenon lamp (100 mW/m²), using MB dye as the target pollutant. The experimental procedures are as follows: 10 mg L⁻¹ MB dye solution was prepared, and the absorbance of the standard solution was measured by a 722-type spectrophotometer, marked as A₀. 100 mL MB working solution was put into a 250 mL glass breaker. Then, a certain amount of S-doped g-C₃N₄ sample was added to this solution. After being stirred for 60 min, the adsorption equilibrium of the dye was obtained, and then the lamp was turned on. The schematic diagram of photocatalytic reactor is displayed in Fig. S1. During the photocatalytic reaction, about 5 mL of suspension was extracted at a specific reaction time. The supernatant was centrifuged and separated. The absorbance of the supernatant was measured and marked as A_i. The MB degradation rate was calculated by the following equation: $Y(\%) = 100(A_0 - A_i)/A_0$. Meanwhile, under identical experimental conditions, random experiments were carried out to verify the repeatability of the experimental results.

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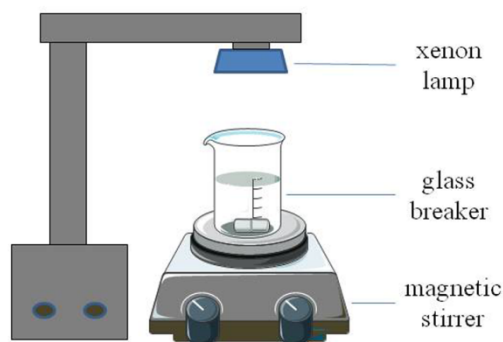


Fig. S-1. Schematic diagram of the photocatalytic reactor.

RSM optimization method

In this work, four parameters were selected as variables, namely initial MB concentration (mg L^{-1}), catalyst dosage (g L^{-1}), solution pH, and reaction time (min), which were labeled as X_1 , X_2 , X_3 , and X_4 , respectively. At the same time, the degradation rate of MB dye was used as the output response (Y). The experimental design, mathematical modeling and response optimization were completed by Design Expert 8.0.7.1 software (Stat-Ease, Inc.). For statistical calculation, the variable X_i was encoded as x_i according to the equation (S1):

$$x_i = \frac{X_i - X_0}{\delta X} \quad (\text{S-1})$$

Where, x_i is the coded value, X_i is the uncoded value, X_0 is the center point value of X_i , and δX represents the step size. In addition, a second-order regression model was used to determine and analyze the response of independent variables. The formula is shown in Equation (S2):

$$Y = \beta_0 + \sum_{i=1}^k \beta_i X_i + \sum_{i=1}^k \beta_{ii} X_i^2 + \sum_i \sum_j \beta_{ij} X_i X_j \quad (\text{S-2})$$

Where, Y is the response value (MB degradation rate), X_i and X_j are the variables that affect the response value, β_i is the first-order regression coefficient, β_{ii} is the second-order regression coefficient, and β_{ij} is the interaction coefficient of two independent variables. Finally, the significance and correlation coefficient of the mathematical model were determined by analysis of variance (ANOVA). The quality and predictive reliability of the polynomial were determined by the correlation coefficient (R^2). The range and level of independent variables for MB photodegradation over S-doped $\text{g-C}_3\text{N}_4$ are shown in Table S-I.

TABLE S-I. Variables and their codes with corresponding real experimental values

Variables	Coded levels				
	-2	-1	0	1	2
X_1 (mg L^{-1})	3	9	15	21	27
X_2 (g L^{-1})	1	2	3	4	5
X_3	3	5	7	9	11
X_4 (min)	30	60	90	120	150

470 *Economic calculation method*

471 *(1) Cost calculation of raw materials*

472 The cost of raw materials mainly included the preparation cost of S-doped g-C₃N₄ and the
473 consumption cost of acid-alkaline regulating reagents, which were the core component of the
474 fixed cost of the photocatalytic process over S-doped g-C₃N₄.

475 The unit preparation cost of S-doped g-C₃N₄ included raw material procurement, energy
476 consumption, and equipment depreciation, and was determined to be 80 CNY kg⁻¹. The
477 preparation cost for treating 1 m³ of MB dye wastewater was calculated as follows: Catalyst
478 cost = catalyst mass (kg) × 80 CNY kg⁻¹.

479 The pH adjustment required sodium hydroxide (or hydrochloric acid), with an industrial
480 cost of approximately 2.5 CNY kg⁻¹. The acid/alkali reagent cost for treating 1 m³ of MB dye
481 wastewater was calculated as follows: acid/alkali reagent cost = acid/alkali reagent dosage
482 (kg) × 2.5 CNY kg⁻¹.

483 *(2) Cost calculation of operation*

484 The operating costs mainly included energy consumption (irradiation and stirring) and
485 labor costs. As the main variable-cost component, operating cost depended directly on the
486 reaction time and treatment scale.

487 Energy consumption in this photocatalytic process primarily arose from the light source
488 and the stirrer. A 300 W xenon lamp was used for irradiation, with an electricity tariff of 0.6
489 CNY kWh⁻¹, and stirring was provided by a 50 W mixer. As the energy consumption increased
490 with reaction time, the energy-consumption cost for treating 1 m³ of methylene blue wastewater
491 was calculated as: energy cost = [light source power (kW) + stirrer power (kW)] × reaction time
492 (h) × 0.6 CNY kWh⁻¹.

493 The labor cost mainly included wastewater monitoring, parameter adjustment, equipment
494 maintenance, and other labor costs. The per capita wage of industrial production was set as
495 6,000 CNY month⁻¹, equivalent to 30 CNY hour⁻¹. Assuming that the operation time was
496 positively correlated with the reaction time, the labor cost for treating 1 m³ MB dye wastewater
497 was calculated as follows: labor cost = reaction time (h) × 30 CNY h⁻¹.

498 *(3) Integrated cost calculation*

499 Combined with raw material cost and operation cost, the integrated cost for treating 1 m³
500 MB dye wastewater could be calculated as follows: integrated cost = (raw material cost +
501 operation cost) = (catalyst cost + acid/alkali reagent cost + energy cost + labor cost).
502

503

RESULTS AND DISCUSSION

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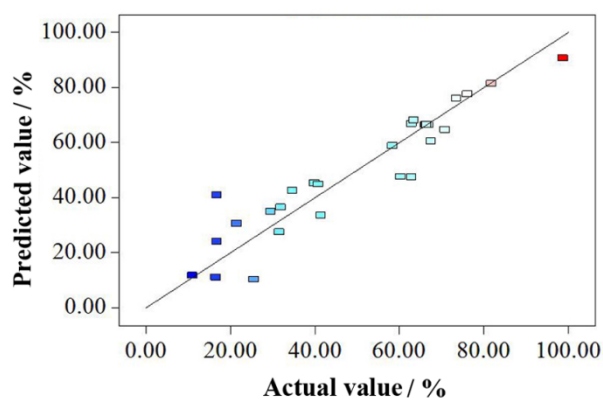
TABLE S-II. Experimental matrix designed by CDD and experimental and predicted values of Y response

runs	X ₁	X ₂	X ₃	X ₄	Y (%)	
					experimental	predicted
1	0.000	0.000	0.000	0.000	66.78	66.39
2	-1.000	-1.000	1.000	1.000	62.91	61.26
3	0.000	0.000	0.000	0.000	66.24	66.39
4	2.000	0.000	0.000	0.000	63.4	60.13
5	-1.000	1.000	-1.000	-1.000	11.02	5.64
6	0.000	2.000	0.000	0.000	16.78	17.38
7	-1.000	1.000	-1.000	1.000	31.93	31.90
8	-2.000	0.000	0.000	0.000	40.84	42.36
9	-1.000	1.000	1.000	1.000	60.29	60.22
10	1.000	1.000	-1.000	1.000	34.64	38.95
11	1.000	-1.000	1.000	1.000	98.72	95.11
12	0.000	0.000	0.000	-2.000	21.47	23.43
13	0.000	0.000	0.000	2.000	81.74	79.24
14	1.000	-1.000	-1.000	1.000	73.5	69.18
15	-1.000	-1.000	-1.000	1.000	58.38	59.48
16	0.000	-2.000	0.000	0.000	76.11	75.89
17	0.000	0.000	0.000	0.000	66.27	63.18
18	0.000	0.000	-2.000	0.000	25.59	25.78
19	0.000	0.000	2.000	0.000	16.76	16.49
20	0.000	0.000	0.000	0.000	66.17	66.36
21	0.000	0.000	0.000	0.000	66.87	66.36
22	1.000	1.000	-1.000	-1.000	16.53	16.59
23	-1.000	1.000	1.000	-1.000	31.56	32.13
24	0.000	0.000	0.000	0.000	66.43	66.36
25	-1.000	-1.000	1.000	-1.000	62.88	61.86
26	-1.000	-1.000	-1.000	-1.000	29.54	29.13
27	1.000	-1.000	1.000	-1.000	70.79	71.00
28	1.000	-1.000	-1.000	-1.000	39.89	40.16
29	1.000	1.000	1.000	1.000	67.48	67.16
30	1.000	1.000	1.000	-1.000	41.42	41.10

506

507 Table S-III. ANOVA analysis for second-order polynomial regression.

Source	Sum of squares	Degree of freedom	Mean square	F value	p value (Prob > F)	
model	13659.70	14	975.69	8.09	0.0001	significant
X ₁	811.77	1	811.77	6.73	0.0204	
X ₂	4277.34	1	4277.34	35.45	< 0.0001	
X ₃	1394.77	1	1394.77	11.56	0.0040	
X ₄	3869.94	1	3869.94	32.07	< 0.0001	
X ₁ X ₂	120.56	1	120.56	1.00	0.3334	
X ₁ X ₃	45.83	1	45.83	0.38	0.5469	
X ₁ X ₄	46.24	1	46.24	0.38	0.5452	
X ₂ X ₃	9.99	1	9.99	0.083	0.7775	
X ₂ X ₄	0.72	1	0.72	0.006	0.9393	
X ₃ X ₄	21.90	1	21.90	0.18	0.6761	
X ₁ ²	168.67	1	168.67	1.40	0.2555	
X ₂ ²	416.88	1	416.88	3.45	0.0828	
X ₃ ²	2862.65	1	2862.65	23.72	0.0002	
X ₄ ²	186.64	1	186.64	1.55	0.2327	
residual	1810.03	15	120.67			
R ²	0.9588					
adjusted R ²	0.9189					



508 Fig. S-2. The plots of predicted values versus actual values.

511 Table S-IV. Comparative analysis data of raw material cost for treating 1 m³ MB wastewater
 512 before and after optimization.

	catalyst cost data		acid-alkaline cost data		net saving (CNY)
	dosage (kg)	cost (CNY)	dosage (kg)	cost (CNY)	
before optimization	5	400	0	0	
after optimization	4.1	328	0.12	0.3	71.7
saving cost	—	72	—	-0.3	

513 Table S-V. Comparative analysis data of operation cost for treating 1 m³ MB wastewater before
 514 and after optimization.

	energy cost data		labor cost data		net saving (CNY)
	energy consumption (kWh)	cost (CNY)	operation time (h)	cost (CNY)	
before optimization	1.050	0.630	3.5	105	
after optimization	0.742	0.445	2.5	75	30.185
saving cost	—	0.185	—	30	

515