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Response Surface Methodology (RSM)-Based Prediction and Optimization of the Fenton Process in Landfill Leachate Decolorization

Anita Maslahati Roudi ¹,*¹, Sultan Salem ²,*¹, Masoud Abedini ^{3,4}, Amin Maslahati ⁵ and Muhammad Imran ⁶

- ¹ Department of Environmental Engineering, Faculty of Civil Engineering, Universiti Teknologi Malaysia, Johor 81310, Malaysia
- ² Department of Economics, Birmingham Business School, University of Birmingham, Edgbaston, Birmingham B15 2TT, UK
- ³ School of Civil Engineering and Architecture, Hainan University, Haikou 570228, China; ma-soud.a877@yahoo.com
- ⁴ School of Civil Engineering, Qingdao University of Technology, Qingdao 266033, China
- ⁵ Faculty of Mechanical Engineering, Qeshm Non-Profit University, Gheshm 79516-15897, Iran; amin.maslahati@yahoo.com
- ⁶ Business Studies Department, Bahria Business School, Bahria University Islamabad, Islamabad 44000, Pakistan; imran.buic@bahria.edu.pk
- * Correspondence: maslahati.anita@gmail.com (A.M.R.); s.salem@bham.ac.uk (S.S.)

Abstract: As an advanced oxidative processes, the Fenton process is receiving popularity as a wastewater treatment technique that can be used for hazardous landfill leachate. The treatment is simple, yet involves complex interactions between the affecting parameters including reaction time, H_2O_2/Fe^{2+} ratio, pH, and iron (II) ion concentration. Hence, the purpose of this present study was to analyze the factors affecting landfill leachate treatment as well as their interaction by means of response surface methodology (RSM) with central composite design. The independent variables were reaction time, H_2O_2/Fe^{2+} ratio, iron (II) ion concentration, and pH, and the dependent variable (response) was color-removal percentage. The optimum treatment conditions for pH, H_2O_2/Fe^{2+} ratio, Fe²⁺ concentration, and reaction time were 8.36, 3.32, 964.95 mg/L, and 50.15 min, respectively. The model predicted 100% color removal in optimum conditions, which was close to that obtained from the experiment (97.68%). In conclusion, the optimized Fenton process using the RSM approach promotes efficient landfill leachate treatment that is even higher than that already reported.

Keywords: landfill leachate; Fenton process; color reduction; RSM; prediction and optimization

1. Introduction

Liquids may become contaminated from dissolved or suspended materials when passing through municipal solid waste landfills. The term used for these liquids is landfill leachate; it is commonly a dark color with strong smell and carries high organic and inorganic loads [1]. There are over 200 organic compounds identified in landfill leachate including aromatic hydrocarbons and cyclic and bicyclic compounds, with each compound having a concentration range from <1 g/L to more than 100 g/L. The contaminant loads can be divided into four groups—heavy metals (Pb²⁺, Ni²⁺, Cd²⁺, Zn²⁺, Cr³⁺, and Cu²⁺), major inorganic ions (NH₄⁺, Fe²⁺, Ca²⁺, Mn²⁺, Na⁺, Cl⁻, K⁺, SO₄⁻², Mg²⁺, and HCO₃⁻), dissolved organic matter (including volatile fatty acids as well as persistent organic matters such as humic compounds), and xenobiotic organic substances derived from chemical and domestic residue (including aromatic hydrocarbons, phenols, and pesticides). Microorganisms are also present in landfill leachate predominated by total and thermo-tolerant coliforms [2–4]. Regardless of its toxicity and recalcitrance, landfill leachate persists in increasing amounts owing to the landfilling methods used for either municipal or industrial solid waste discharge [5].



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Current approaches employed in leachate treatment are mostly dependent on biological techniques. However, such approaches are considered impractical due to extremely variable organic loads and leachate flows as well as the requirements of vast construction areas. Furthermore, treatment efficiency could be reduced due to the refractory features of the leachate [6]. Other approaches, including adsorption, also require special reactors and work only against specific contaminants [7,8]. Alternatively, combined physicochemical processes can be used on the leachate to degrade and mineralize the refractory contaminants [9]. An example of this is the recently popular treatment method, advanced oxidative processes (AOP_s), that can work against a wide range of resistant contaminants. The Fenton process, one of AOPs, works by exploiting the vigorous oxidation potential of •OH radicals which results in the mineralization of resistant contaminants and biodegradability enhancement allowing for further treatment using biological techniques. The •OH radicals are produced concomitant to the catalysis reaction of hydrogen peroxide using ferrous ions [8]. The treatment system itself consists of pH adjustment, oxidation reaction, neutralization, coagulation, and settlement [10,11]. Nonetheless, despite its high removal efficiency, previous work employing Fenton reagents in leachate treatment has reported unsatisfying results, where a post-treatment was required for its application [12].

Based on a previous report [13], pH, experimental time, the amount of Fe^{2+} , and the H_2O_2 : Fe^{2+} ratio are significant to the outcome of oxidative leachate treatment. Optimization of the aforementioned parameters could be a solution to overcoming the weaknesses in the Fenton process. In this light, advanced statistical design is commonly employed for modelling, optimization, and characterization of a particular treatment [14]. Herein, we employed response surface methodology (RSM) which is a combination of mathematical and statistical techniques to study correlated factors in leachate treatment process and their optimization [15]. In comparison to conventional optimization methods, the RSM could provide much more information from only a small number of experiments performed [16]. In this study, the optimum conditions (reaction time, pH, and H_2O_2 :Fe²⁺ ratio, as well as Fe²⁺ concentration) for landfill leachate decolorization by the Fenton process were determined. Using the color-removal percentage is a common parameter to evaluate the success of the landfill leachate treatment [17–19]. The experimental runs were prepared based on the suggestion from central composite design (CCD) [20].

2. Materials and Methods

2.1. Sampling and Characterization of the Landfill Leachate

In the Jeram Sanitary Landfill the leachate produced from landfill is pumped into a treatment system that consists of three main stages—primary stage, secondary stage, and tertiary stage. During the primary stage the leachate is passed through three equalization ponds (also referred to as sequential batch reactors). The same process is repeated in the secondary stage but with only two sequential batch reactors. Once the secondary stage is completed the remaining liquid is subjected to physical and chemical treatments during the tertiary stage to ensure compliance with standards. In this work, leachate samples were obtained from a landfill near an oil palm plantation at Jeram, Kuala Selangor, Malaysia. Priorly-cleaned two-gallon plastic containers were used to carry the leachate sample collected from beneath the top surface. Each container was labelled according to the sampling location and date. Before being capped, air bubbles in the sample were removed. All samples were brought to the laboratory and kept at 4 °C. Sieve analysis (gradation test with pore size 100 and 150 um), $0.45 \,\mu$ m Whatman filter glass paper and a centrifuge device (Kubota 2420 model) were used to exclude large particles and debris before the experiment. The results of the leachate samples showed the concentration of organic compounds, expressed as COD, were 10,516 mg/L and the ammoniacal (NH₃-N) concentration in the leachate studied was about 15 mg/L. Characteristics of the collected leachate sample are presented in Table 1.

Test Parameter	Unit	Result (mg/L)
pН	-	7.5
Temperature	°C	40
COD	mg/L	10,516
Total suspended solid	mg/L	810
Oil and grease	mg/L	9.5
Zinc as Zn	mg/L	2.48
Iron as Fe	mg/L	4.8
Chromium as Cr	mg/L	0.15
Arsenic as As	mg/L	0.17
Aluminium as Al	mg/L	20
Barium as Ba	mg/L	2.75
Formaldehyde	mg/L	1.9
Ammonia nitrogen	mg/L	15
Color original pH	ADMI	>500
Color adjusted to pH 7.0	ADMI	>500

Table 1. Characteristics of the leachate sample collected from Jeram Landf

2.2. Fenton Experiments

To obtain the data on effective variables and their proper range, Design Expert software was employed to produce a full factorial design effect plot. Investigation on possible combinations was carried out under the full factorial approach that is required to generate a probability model displaying interactions between *k* design variables. As many as four steps and 20 runs per experimental set were involved in the screening phase (total runs: 80 with 2^k design, where *k* represents the number of variables). As suggested previously [16], in this present study, 2^k full factorial design was employed, where responses were measured from all combinations of the experimental factor levels. Preliminary four-steps full factorial design has been presented (Table 2). Plots were generated to determine each variable and to observe its interaction against the response variables. The ratio of H₂O₂:Fe²⁺ (2, 5, 6, and 10) was assigned as the variable factor in the selection of full factorial design. The concentration variation of Fe²⁺ was prepared as per the suggestions of previous studies [15,21]. The following are the operated runs at each Fe²⁺ concentration.

Run	Factor A: pH	Factor B: Time (min)	Factor C: H ₂ O ₂ :Fe ²⁺ Ratio
1	6	105	6
2	3	105	6
3	6	105	6
4	9	180	10
5	6	60	6
6	3	180	2
7	9	105	6
8	3	180	10
9	6	60	6
10	3	30	2
11	6	105	5
12	6	105	6
13	6	105	6
14	6	105	5
15	9	30	2
16	6	105	6
17	9	180	2
18	3	30	10
19	9	30	10
20	6	105	6

Table 2. Preliminary full factor design for color removal resulting from the Fenton process.

Step 1: 500 mg/L Fe²⁺ Step 2: 4000 mg/L Fe²⁺ Step 3: 1000 mg/L Fe²⁺ Step 4: 2000 mg/L Fe²⁺

2.3. Experimental Design and Statistical Model

In this procedure, contact time (A), Fe²⁺ concentration (B), pH (C), as well as ratio of H_2O_2 :Fe²⁺ (D) were assigned as independent variables determined using the RSM. As for the responses, overall color removal (Y_1) was selected. Effect plots were generated in R (R core team) software version 3.2.2 based on an assessment of collective responses from 80 runs, suggesting the effective factors. Exclusion was then applied on factors that were not effective. Hence, the new design consisted of 44 runs. CCD was used in the RSM model to observe the interaction between the independent variables against Fenton-treatmentinduced color removal. The levels of independent variables and experimental range employed toward Fenton-process-based leachate treatment have been presented (Table 3). Meanwhile, the yielded responses (Y_1) and experimental values (A, B, C, and D), along with variable-encoded experimental data are given in Table 4. Each response represents a function of the treatment time (A), Fe^{2+} concentration (B), pH (C), and H_2O_2 : Fe^{2+} ratio (D). Important terms of the interaction effects (AB, AC, AD, BC, BD, and CD), the first order (A, B, C, and D), and the second-order (A^2 , B^2 , C^2 , and D^2) are included. The approximate ranges of the variables were 5–60 min, 500–1500 mg/L, pH 3–9, and 2–10 for treatment time, Fe²⁺ concentration, initial pH level, and H₂O₂:Fe²⁺ ratio, respectively. Collectively, there were 44 experiments designed using 2k; 8 axial (star) points, 16 factorial points, and 20 replicates at the center point. The chosen independent factors were coded using Equation (1) [22,23].

$$X_i = \frac{(X_i - X_0)}{\Delta X} \tag{1}$$

where X_i represents a dimensionless coded value of the *i*th independent variable, while X_0 and ΔX are the center point value of X_i and step change value, respectively. To determine the correlation between the variables (X) and the responses (Y), a quadratic model was applied (Equation (2))—an adequate (second order) model. A linear model predicting the responses from entire experimental fields was also generated—approximating the interaction between the decolorization and the four independent variables [8]:

$$Y = \beta_0 + \sum_{j=1}^k \beta_j X_j + \sum_{j=1}^k \beta_{jj} X_j^2 + \sum_{i=1}^{k-1} \sum_{j=2}^k \beta_{ij} X_i X_j + e$$
(2)

where b_i , b_{ii} , and b_{ij} represent the coefficient of the related regression coefficients. Obtained data were analyzed statistically using ANOVA operated in Minitab software. Coefficient of determination (R²) and adjusted R² represent the model's fitting quality. *F*-tests were further used to assess the significance of linear and quadratic terms. *P*-value was used to select the final subset of variables at a confidence level of 95%. Predictive ability of the proposed model was judged based on R² prediction coefficient along with predicted residual error sum of squares (PRESS). Relationships between responses and experimental levels of each tested variable were illustrated in contour and surface plots based on the fitted polynomial equation. Preparation of the 3D response plots and contour plots was based on a customized program using design expert software. The optimum value of each independent variable was tested in a batch Fenton experiment for validation.

Run	Time (min)	Fe ²⁺ Concentration	pH	H ₂ O ₂ :Fe ²⁺ Ratio
1	32.5	1000	6	6
2	46.25	1250	7.5	8
3	32.5	1000	6	6
4	46.25	750	7.5	4
5	46.25	1250	4.5	4
6	18.75	1250	7.5	8
7	32.5	1000	6	6
8	32.5	1000	6	6
9	18.75	750	7.5	4
10	46.25	1250	7.5	4
11	32.5	1000	6	6
12	32.5	1000	6	6
13	18.75	750	4.5	4
14	18.75	750	7.5	8
15	32.5	1000	6	6
16	46.25	1250	4.5	8
17	32.5	1000	6	6
18	32.5	1000	6	6
19	18.75	1250	4.5	4
20	32.5	1000	6	6
21	32.5	1000	6	6
22	18.75	1250	7.5	4
23	46.25	750	7.5	8
24	46.25	750	4.5	4
25	46.25	750	4.5	8
26	32.5	1000	6	6
27	18.75	750	4.5	8
28	18.75	1250	4.5	8
29	32.5	1000	6	6
30	32.5	1000	6	2
31	32.5	1000	3	6
32	32.5	1000	6	6
33	32.5	1000	6	10
34	32.5	1500	6	6
35	32.5	500	6	6
36	32.5	1000	6	6
37	32.5	1000	6	6
38	32.5	1000	6	6
39	32.5	1000	6	6
40	32.5	1000	6	6
41	32.5	1000	9	6
42	32.5 F	1000	6	6
43	5	1000	6	6
44	60	1000	0	0

 Table 3. Central composite experiments design.

 Table 4. Experimental range and levels used for the independent variables.

			Coded Levels						
		-α	-1	0	1	α			
Variable	Symbol			Real Values					
Time	X ₁	5	18.75	32.5	46.25	60			
Fe ²⁺	X ₂	500	750	1000	1250	1500			
pН	X ₃	3	4.5	6	7.5	9			
Ratio	X_4	2	4	6	8	10			

The influential factors such as initial pH, H_2O_2 :Fe²⁺ ratio, and Fe²⁺ concentration, along with their ranges were chosen according to published studies [24–27], as shown in Table 3. The coded levels and the factor values used in the experiments are presented in Table 4. The batch experiment was conducted in a 1 L glass reactor using jar-test equipment with flat stirring vanes (25 °C; 1 atm). The 300 mL leachate sample was poured onto the glass reactor, followed by pH adjustment (3, 4.5, 6, 7.5, and 9) using H₂SO₄ 95–97%. Ferrous sulphate (FeSO₄.7H₂O) powder (500, 750, 1000, 1250, and 1500 mg/L) was then added to the reactor and mixed for 5 min until homogenous. Next, to initiate the Fenton reaction, H₂O₂ 30% (*w/w*) was added until the H₂O₂:Fe²⁺ ratio reached the desired amount (2, 4, 6, 8, and 10). The procedure was followed by a rapid mixing (250 rpm) for 80 s in a jar-test, before the rotation speed slowed down to 50 rpm for predetermined contact time (5, 18.75, 32.5, 46.25, and 60 min). Once completed, the sample was filtered at the same intervals and subsequently neutralized using NaOH until pH 7.5–8.0. The sample was left to allow the formation of precipitate for 1 h. The supernatant was measured for color removal using the following equation [28]:

$$(COD_{initial} - COD_{final}/COD_{initial}) \times 100$$
 (3)

3. Results and Discussions

3.1. Preliminary Investigation

For this part of study, there were four steps that each included 20 runs. These runs were performed for screening. In total, there were 80 runs conducted as provisioned by the full factorial method in the laboratory during the preliminary investigation. A number of runs were applied on Fe^{2+} concentration variation based on the predetermined H_2O_2 : Fe^{2+} ratios. As the outcome, the stage of investigation determined effective variables along with their appropriate ranges. The aim of choosing this parameter was to reduce the number of runs. As a consequence, this phase was able to determine the effective variables and their adequate ranges. In order to assess each variable and how it affects the response variables, effect plots were prepared based on the results. The orders of runs are shown as follows. To assess the effect of each variable against the decolorization, effect plots were generated and presented (Figure 1). The order of runs is exhibited below.

The range of solution pH appears to influence the color removal, which is in contrast with the reaction time that yielded insignificant effect.

This means with the 500 mg/L Fen²⁺ the color removal would be at pH (acidic range) and the ratio (alkaline range) had a significant effect but contact time has little effect. For color removal with the 1000 mg/L Fe²⁺, the ratio (alkaline range) had a significant effect, but contact time did not. It also revealed that ratio had more effect on the process in comparison to the initial pH. The color removal with the 2000 mg/L Fe²⁺, pH (alkaline range) had a considerable influence, although the reaction time and ratio did not have any great effect.

3.2. Primary Investigation

The experimental data obtained through 44 different observations have been normalized. Each independent variable had a variation ranged between $-\alpha$, -1, 0, and +1, $+\alpha$ that was determined as per suggestions from the preliminary study along with published reports. This coding scheme was massively employed in fitting regression models, where the coded variables would fall between $+\alpha$ (maximum value) and $-\alpha$ (minimum value) [24,29]. The summary of adopted configurations in each CCD are presented in Table 5.



Figure 1. Effect plots of the independent variables on color removal for (**a**) pH, (**b**) time, (**c**) H_2O_2 : Fe²⁺ ratio.

Run	Time (A)	Fe ²⁺ (B)	pH (C)	H ₂ O ₂ :Fe ²⁺ Ratio (D)	Time (min)	Fe ²⁺ mg/L	pН	Ratio
1	1	-1	-1	-1	46.25	750	4.5	4
2	1	1	1	1	46.25	1250	7.5	8
3	0	0	0	0	32.5	1000	6	6
4	0	0	0	0	32.5	1000	6	6
5	0	0	0	0	32.5	1000	6	6
6	0	0	0	0	32.5	1000	6	6
7	0	0	0	0	32.5	1000	6	6
8	0	0	0	0	32.5	1000	6	6
9	0	0	0	0	32.5	1000	6	6
10	0	0	0	0	32.5	1000	6	6
11	-1	-1	1	-1	18.75	750	7.5	4
12	-1	-1	-1	1	18.75	750	4.5	8
13	0	0	0	0	32.5	1000	6	6
14	1	1	-1	-1	46.25	1250	4.5	4
15	0	0	0	0	32.5	1000	6	6
16	-1	1	1	1	18.75	1250	7.5	8
17	0	0	0	0	32.5	1000	6	6

Table 5. Coded variable of central composite experiments design.

Run	Time (A)	Fe ²⁺ (B)	pH (C)	H ₂ O ₂ :Fe ²⁺ Ratio (D)	Time (min)	Fe ²⁺ mg/L	рН	Ratio
18	0	0	0	0	32.5	1000	6	6
19	0	0	0	0	32.5	1000	6	6
20	1	1	1	-1	46.25	1250	7.5	4
21	0	0	0	0	32.5	1000	6	6
22	0	0	0	0	32.5	1000	6	6
23	1	1	-1	1	46.25	1250	4.5	8
24	-1	1	-1	-1	18.75	1250	4.5	4
25	-1	-1	1	1	18.75	750	7.5	8
26	-1	1	1	-1	18.75	1250	7.5	4
27	0	0	0	0	32.5	1000	6	6
28	1	-1	1	1	46.25	750	7.5	8
29	0	0	0	0	32.5	1000	6	6
30	1	-1	-1	1	46.25	750	4.5	8
31	-1	1	-1	1	18.75	1250	4.5	8
32	0	0	0	0	32.5	1000	6	6
33	0	0	0	0	32.5	1000	6	6
34	-1	-1	-1	-1	18.75	750	4.5	4
35	0	0	0	0	32.5	1000	6	6
36	1	-1	1	-1	46.25	750	7.5	4
37	0	$-\alpha$	0	0	32.5	500	6	6
38	α	0	0	0	60	1000	6	6
39	0	α	0	0	32.5	1500	6	6
40	0	0	$-\alpha$	0	32.5	1000	3	6
41	$-\alpha$	0	0	0	5	1000	6	6
42	0	0	α	0	32.5	1000	9	6
43	0	0	0	$-\alpha$	32.5	1000	6	2
44	0	0	0	α	32.5	1000	6	10

Table 5. Cont.

3.3. Color Removal Efficiency Using the RSM Method

This part evaluated the color removal efficiency of landfill leachate treatment relying on the Fenton process. Table 6 shows the color-removal percentage in accordance with the experimental and predicted data using the RSM. The maximum color removal was obtained in run 36 with 95% removal and the minimum color removal in run 31 with 13.4% removal. The predicted data were obtained from Equation (4). The RSM method revealed an overall prediction error of 4.3%. This suggests the suitability of the proposed empirical method in forecasting the color removal.

Table 6. Experimental design and results for color removal.

Run	Time	Fe	pН	Ratio	Color Removal%	Predicted Color R%	Error%
1	46.25	750	4.5	4	48.50	41.481	14.470
2	46.25	1250	7.5	8	42.00	45.029	-7.212
3	32.5	1000	6	6	65.73	54.903	16.470
4	32.5	1000	6	6	68.00	54.903	19.259
5	32.5	1000	6	6	63.90	54.903	14.078
6	32.5	1000	6	6	61.63	54.903	10.914
7	32.5	1000	6	6	67.14	54.903	18.225
8	32.5	1000	6	6	68.31	54.903	19.625
9	32.5	1000	6	6	65.40	54.903	16.049
10	32.5	1000	6	6	63.70	54.903	13.809
11	18.75	750	7.5	4	84.59	77.074	8.884
12	18.75	750	4.5	8	35.80	36.651	-2.379
13	32.5	1000	6	6	59.70	54.903	8.034
14	46.25	1250	4.5	4	48.16	49.778	-3.360
15	32.5	1000	6	6	65.40	54.903	16.049
16	18.75	1250	7.5	8	16.00	12.718	20.509
17	32.5	1000	6	6	61.50	54.903	10.725

Run	Time	Fe	pН	Ratio	Color Removal%	Predicted Color R%	Error%
18	32.5	1000	6	6	55.00	54.903	0.175
19	32.5	1000	6	6	61.00	54.903	9.994
20	46.25	1250	7.5	4	85.00	73.863	13.101
21	32.5	1000	6	6	53.70	54.903	-2.241
22	32.5	1000	6	6	48.67	54.903	-12.807
23	46.25	1250	4.5	8	39.17	36.327	7.256
24	18.75	1250	4.5	4	30.00	23.031	23.229
25	18.75	750	7.5	8	72.69	64.769	10.896
26	18.75	1250	7.5	4	43.72	40.582	7.175
27	32.5	1000	6	6	48.30	54.903	-13.672
28	46.25	750	7.5	8	81.65	78.243	4.172
29	32.5	1000	6	6	49.00	54.903	-12.048
30	46.25	750	4.5	8	46.70	43.591	6.656
31	18.75	1250	4.5	8	13.40	10.550605	21.26414
32	32.5	1000	6	6	50.15	54.903	-9.478
33	32.5	1000	6	6	50.36	54.903	-9.022
34	18.75	750	4.5	4	42.84	33.572	21.633
35	32.5	1000	6	6	56.00	54.903	1.957
36	46.25	750	7.5	4	95	91.517	3.665
37	32.5	500	6	6	57.00	65.556	-15.010
38	60	1000	6	6	58.60	61.401	-4.780
39	32.5	1500	6	6	22.80	21.801	4.381
40	32.5	1000	3	6	22.30	23.836	-6.890
41	5	1000	6	6	15.80	21.180	-34.056
42	32.5	1000	9	6	70.00	76.040	-8.628
43	32.5	1000	6	2	57.00	67.665	-18.711
44	32.5	1000	6	10	45.00	41.9103	6.866

Table 6. Cont.

3.4. Regression Models and Statistical Analysis

The polynomial model representing the response variable (color removal) was constructed based on the data presented in Table 4. Further, these data were used to prepare a regression model; the *Y* response was defined as a function of contact time (*A*), Fe²⁺ concentration (*B*), pH (*C*), and the H₂O₂:Fe²⁺ ratio (*D*). The proposed formula is as follows:

 $Y(Color Removal) = 0.109 \times A + 0.152 \times B + 37.739 \times C + 12.626 \times D + 0.00137 \times (AB) + 0.0792 \times (AC)$ $-0.00882 \times (AD) - 0.0173 \times (BC) - 0.00778 \times (BD) - 1.28 \times (CD) - 0.018 \times A^{2}$ $-0.00004 \times B^{2} - 0.551 \times C^{2} - 0.00723 \times D^{2} - 180.41$ (4)

where *A* represents contact time, *B* is the Fe²⁺ concentration, *C* is the pH, and *D* is the H_2O_2 :Fe²⁺ ratio, with *Y* as the response (in this case, it is % color removal). The scope of the generated formula is to determine the efficiency of color removal from landfill leachate by using the Fenton process. In addition, the empirical formula was developed to remove the turbidity of leachate.

A significant effect of the Fe²⁺ concentration, pH, time, H₂O₂:Fe²⁺ratio, and interactions among the variables was observed. Table 7 illustrates the model *F*-value of 18.53 suggesting statistical significance. It could be a 0.01% chance that noise is responsible for a "Model *F*-Value" of this size. In this investigation, *A*, *B*, *C*, *D*, *AB*, *BC*, *BD*, *CD*, *A*², and *B*² are statistically meaningful derived from their values of Prob > F that are less than 0.05. Meanwhile, when the values are higher than 0.1, the model terms are not considered significant. Model reduction could contribute to an improvement when many insignificant model terms were found (excluding the required terms for hierarchy support). The "Lack of Fit *F*-value" of 0.85 suggests that the lack of fit is relatively insignificant to the pure error. This large "Lack of Fit *F*-value" has 58.12% chance to occur. Note that insignificant lack of fit is preferred.

Fitting quality of a model could be judged by the coefficient of determination (\mathbb{R}^2). ANOVA results in this study present a high \mathbb{R}^2 value (0.9026) that is close to the adjusted \mathbb{R}^2 (0.8539) implying the applicability of the quadratic model. The closeness between \mathbb{R}^2 and adjusted \mathbb{R}^2 values suggests that the all terms included in the model were significant [30]. A model could not be assessed solely by \mathbb{R}^2 , where a model with high \mathbb{R}^2 may still yield poor prediction. Thus, the predictability of a regression model could be assessed by \mathbb{R}^2 prediction, derived from the prediction error sum of squares (PRESS) [24,31]. Generally, our model was found to have better predictability according to the stated criterion when compared with others [26,32]. Table 8 presents a "Pred R-Squared" of 0.7156 which is in line with the adjusted R² of 0.8539. Moreover, Adeq Precision was taken to measure the signal-to-noise ratio, where the value higher 4 is desirable. Herein, Adeq Precision of 19.156 was obtained indicating an adequate signal.

 Table 7. Analysis of variance for color removal.

Source	Sum of Squares	df	Mean Square	F Value	<i>p</i> -Value Prob > F	Remark
Model	12,374.6	14	883.9	18.53	< 0.0001	Significant
A-Time	2257	1	2257	47.32	< 0.0001	Significant
B-Fe	2789	1	2789	58.47	< 0.0001	Significant
С-рН	4042.49	1	4042.49	84.75	< 0.0001	Significant
D-Ratio	993.31	1	993.31	20.82	< 0.0001	Significant
AB	354.19	1	354.19	7.43	0.011	Significant
AC	42.77	1	42.77	0.9	0.3518	Insignificant
AD	0.94	1	0.94	0.02	0.8893	Insignificant
BC	677.3	1	677.3	14.2	0.0008	Significant
BD	241.96	1	241.96	5.07	0.0323	Significant
CD	236.7	1	236.7	4.96	0.0341	Significant
A^2	397.23	1	397.23	8.33	0.0074	Significant
B^2	258.04	1	258.04	5.41	0.0275	Significant
C^2	50.58	1	50.58	1.06	0.3119	Insignificant
D^2	0.027	1	0.027	$5.76 imes10^{-4}$	0.981	Insignificant
Residual	1335.55	28	47.7			-
Lack of Fit	383.72	9	42.64	0.85	0.5812	Insignificant
Pure Error	951.83	19	50.1			-
Cor Total	14,685.97	43				

Table 8. Regression statistics.

Regression Parameter	Magnitude	Regression Parameter	Magnitude
Std. Dev.	6.91	R-Squared	0.9026
Mean	53.55	Adj R-Squared	0.8539
C.V. %	12.9	Pred R-Squared	0.7156
PRESS	3898.94	Adeq Precision	19.156

A popular combination technique of the used procedures in this study, the stepwise regression method, allows simultaneous addition or deletion of regressors, where only variables with statistical significance (p < 0.05) are be included. Hence, the statistically insignificant A:C, A:D, C², and D² coefficients were excluded from the model. As a consequence, only interactions between A, B, C, D, AB, AC, AD, BC, BD, CD, A², B², C², and D² were included to the model. Model reduction was then carried out using a new ANOVA. Reduced second-order equation of the model and some other statistical parameters could be observed in Table 9. The Model *F*-value was 27.48 suggesting the statistical significance of the model.

In this case, there was merely a 0.01% chance that the noise generated such "Model *F*-Value". Moreover, the significance of the model was exhibited by values of "Prob > F" less than 0.05. It can be concluded that A, B, C, D, AB, BC, BD, CD, A², and B² are considered significant terms in the model. The "Lack of Fit *F*-value" of 0.73 suggests that the insignificance of the lack of fit was relative to pure error with a 71.18% chance it was generated from the noise.

Source	Sum of Squares	Degree of Freedom	Mean Square	<i>F</i> Value	<i>p-</i> Value Prob > F	Remark
Model	12,280.180	10	1228.018	27.480	< 0.0001	Significant
A-Time	2256.996	1	2256.996	50.507	< 0.0001	Significant
B-Fe	2789.001	1	2789.001	62.412	< 0.0001	Significant
C-pH	4042.491	1	4042.491	90.463	< 0.0001	Significant
D-Ratio	993.306	1	993.306	22.228	< 0.0001	Significant
AB	354.192	1	354.192	7.926	0.0083	Significant
BC	677.3006	1	677.3006	15.156	0.0005	Significant
BD	241.958	1	241.958	5.414	0.0265	Significant
CD	236.698	1	236.698	5.296	0.0280	Significant
A^2	405.445	1	405.445	9.073	0.0050	Significant
B^2	264.583	1	264.583	5.920	0.0207	Significant
Residual	1429.966	32	44.686			Ū
Lack of Fit	478.135	13	36.779	0.734	0.7118	Insignificant
Pure Error	951.831	19	50.096			Ũ
Cor Total	14,685.970	43				

Table 9. Analysis of variance for color removal (reduced model).

High \mathbb{R}^2 values show the high significance of the model, indicating the reliability of the polynomial equation. Furthermore, all of the screened terms had significant contribution to the proposed model (p < 0.05). The "Pred R-Squared" of 0.8043 was in reasonable agreement with the "Adj R-Squared" of 0.8631. "Adeq Precision" measures the signal to noise ratio. A ratio greater than 4 is desirable. Adeq Precision of 22.852 indicates a proportional signal. This model could be utilized to direct the design space, as shown in Table 10.

Table 10. Regression statistics (reduced).

Regression Parameter	Magnitude	Regression Parameter	Magnitude
Std. Dev.	6.684	R-Squared	0.895
Mean	53.552	Adj R-Squared	0.863
C.V. %	12.482	Pred R-Squared	0.804
PRESS	2683.033	Adeq Precision	22.852

The plot constructed according to the experimental results versus the predicted data is presented in Figure 2. It shows that most data points were well-distributed around the straight-line X = Y in a narrow area. The plot also has $R^2 = 0.87999$ suggesting good reliability of the model. Therefore, this model could be utilized to direct the design space.

An analysis was conducted for the assumption of normality by preparing the plot of studentized residuals vs. the normal % of probability (Figure 3), revealing a satisfactory model towards color removal.

3.4.1. Perturbation Plot for Color Removal

A perturbation plot was constructed for deeper identification of the most sensitive factors for leachate treatment. In this part, the perturbation plot for color removal was used to assess the response behavior as a result of deviation from the center point while the other factors were maintained constant. In the perturbation plot (Figure 4), the increase in reaction time (A) and pH (C) promoted higher color-removal percentage; therefore, the reaction time (A) and pH (C) yielded the most noteworthy positive effect on the color response. Increase in Fe²⁺ concentration (B) along with the ratio of H₂O₂:Fe²⁺ (D) decreased the color-removal percentage suggesting the negative effects on the decolorization.



Figure 2. Predicted vs. actual values plot for color removal.



Figure 3. The normal % probability and studentized residuals.



Figure 4. Perturbation plot for color removal.

3.4.2. Three-Dimensional (3D) Response Surfaces and Contour Plots for COD Removal

To obtain graphical illustrations of the regression equation the 3D response surface plot and 2D contour map were constructed in Design Expert 7.0 software analysis. The first-order effects, six interaction effects (AB, AC, AD, BC, BD, and CD) and four second-order effects (A^2 , B^2 , C^2 , and D^2) were evaluated according to Equation (4).

Variation from 5 to 60 min reaction time was applied to determine its effect on Fenton process-induced color removal. It also revealed that Fenton efficiency is negatively affected by the augmentation of Fe^{2+} concentration (Figure 5). Generally, the uptake of organic contaminants could be enhanced by decreasing concentration of the iron salt, and similarly, by decreasing the reaction time. Nonetheless, the increase in color removal was probably too minor in the presence of high iron salt concentration. According to previously published works, further addition of Fe^{2+} could induce self-inhibition of the •OH radical which consequently reduce the degradation of organic contaminants [28,33,34]:



 $OH^{\bullet} + Fe^{2+} \rightarrow Fe^{3+} + OH^{-}$ (5)

Figure 5. Role of reaction time and Fe²⁺ concentration in leachate decolorization depicted in 3D response surfaces and contour plot.

The influence of pH against the leachate decolorization is exhibited in Figure 6, where the system was varied with pH ranged from 3, 4.5, 6, 7.5 to 9. A decrease in color removal efficiency was observed at pH lower than 7.5, ascribed to the reduction of hydroxyl radical population concomitant to Fe^{2+} complexation [35]. Similarly, the formation of Fe^{2+} complex is also responsible for the decrease of removal efficiency at pH > 7.5. This is in line with a previously reported study suggesting the reduction of oxidation potential of the •OH radicals when the system pH is elevated [35]. Moreover, the leachate that was originally black in color turned into light brown at higher and lower pH levels. The results also show that reaction time has a positive correlation with color removal. The equilibrium was reached after 46.25 min reaction. Afterward, the significant color removal no longer occurred. Reaction time to reach equilibrium varied from 30 min up to 3 h [36,37].



Figure 6. Role of reaction time and pH in leachate decolorization depicted in 3D response surfaces and contour plot.

Figure 7 indicates the increase in decolorization efficiency as a function of time during oxidation. The obtained data demonstrate that the landfill leachate was instantly decolorized during the Fenton process. Commonly, organic contaminants were degraded as soon as 46.25 min reaction time. Afterwards, the decolorization was negligible. Numerous foams were formed on the leachate surface with the proceeding of the oxidation. Bicarbonate ions originating from the leachate were converted into carbonic acid when pH level was reduced to acidic range, which consequently dissolved in the liquid or released it to the air as CO₂. Organic matter present in the leachate was dramatically converted to oxidized by-products as a result of the Fenton process. During the early state of oxidation reactions, the conversion occurs both partially and fully. With the abundance of hydroxyl radicals, the full conversion is expected. Nonetheless, as the time elapses the reaction would only rely only on residual hydroxyl radicals that consequently cause the partial conversion.

The ratio of H_2O_2 :Fe²⁺ was set at a variation from 2, 4, 6, 8, to 10 in order to investigate its effect against the color removal. Both hydrogen peroxide and iron are essential in the Fenton process, especially in regard of the treatment efficiency and operation cost. The efficiency could be obtained by elevating the ratio of H_2O_2 :Fe²⁺ from 2 to 4, whereas further increment led to lower removal percentage. This phenomenon could be associated with the reaction mechanism of the Fenton process as reported previously [36]. At low H_2O_2 :Fe²⁺ ratio, second-pseudo-order kinetics governs the reaction rate until the stoichiometry ratio of 2Fe (II) \cong H_2O_2 is reached. However, when the H_2O_2 :Fe²⁺ ratio goes up, the oxidative reaction is more dependent to zero-order kinetics. Higher H_2O_2 :Fe²⁺ ratios result in the alteration of the mechanism, where it would be no longer affected by hydrogen peroxide concentration [36].



Figure 7. Role of reaction time and H_2O_2 :Fe²⁺ratio in leachate decolorization depicted in 3D response surfaces and contour plot.

As mentioned earlier, due to the importance of Fe^{2+} and H_2O_2 in the Fenton process, the H_2O_2 : Fe^{2+} ratio used in the treatment required optimization. The oxidation of organic substances in the leachate by hydroxyl radicals produced from a reaction between H_2O_2 and iron followed the reaction below [38]:

$$OH^{\bullet}$$
 + organic substances \rightarrow oxidation products (6)

The optimization of H₂O₂ and Fe²⁺ aimed to avoid the excessiveness that could cause unwanted [•]OH radical-scavenging reactions. Thus, H₂O₂:Fe²⁺ ratios used in the batch experiment were varied from 2, 4, 6, 8, to 10. The optimum Fe²⁺ concentration and H₂O₂:Fe²⁺ ratio was revealed to be 750 mg/L and 4, respectively, where the color removal efficiency reached 95% (Figure 8). A higher ratio of H₂O₂:Fe²⁺ could lead to reduced efficiency of the color removal, which is ascribed to the scavenging activity of H₂O₂ against hydroxyl radicals [38]:

$$H_2O_2 + OH^{\bullet} \rightarrow HO^{\bullet}_2 + H_2O \tag{7}$$



Figure 8. Role of Fe^{2+} concentration and H_2O_2 : Fe^{2+} ratio in leachate decolorization depicted in 3D response surfaces and contour plot.

The foregoing reaction induces the generation of the hydroperoxyl radical (HO₂); a species with significantly less oxidizing potential than that of hydroxyl radical. Further-

more, auto-decomposition of H_2O_2 into water along with oxygen and the recombination of •OH radicals could occur at the excess of H_2O_2 . Meanwhile, when the ratio is set below 4, the scavenging activity would be inhibited by the excess of Fe³⁺ species that consequently reduces the color removal efficiency (Equation (8)). It appears that Fe²⁺ could compete with organic contaminants in the hydroxyl-radical-induced oxidation. Hence, Fe²⁺ digests hydroxyl radicals (Equation (8)), leading to the reduction of the treatment efficiency.

$$Fe^{2+} + OH^{\bullet} \rightarrow Fe^{3+} + OH^{-}$$
(8)

As a consequence, more addition of Fe²⁺ results in less availability of $^{\bullet}$ OH radicals to act in oxidizing the organic contaminants. Additionally, Fe³⁺ may form a reaction with hydrogen peroxide (Equation (8)) to yield Fe²⁺ and HO₂ in the leachate. Taken altogether, the ratio of 4 is assigned as the optimum ratio. In this study, the obtained optimum ratio was distinct to that reported. For instance, a 20:1 ratio of H₂O₂:Fe²⁺ yielded an optimum result in a raw leachate treatment [39]. Meanwhile, another previous study revealed the optimum H₂O₂:Fe² ratio was 1.5 [40]. Disagreement on optimum H₂O₂:Fe² ratio between published research and our work could be attributed to different characteristics of the landfill leachate.

$$H_2O_2 + Fe^{3+} \to Fe^{2+} + HO^{\bullet}_2 + H^+$$
 (9)

This study revealed that the highest removal efficiency was reached with Fe^{2+} concentration of 750 mg/L. More addition of Fe^{2+} concentration could be associated with the increase in the decolorization until the optimum concentration was reached. As explained earlier, competition between Fe^{2+} and organic substances might occur leading to inefficient decolorization. Hence, excessive amounts of Fe^{2+} should be avoided.

Interaction between iron (II) ion concentration as well as initial pH of the system against the decolorization is presented in Figure 9. The interaction suggests that modification of Fe²⁺ and H₂O₂ is dependent on the pH level, where all of them are regressor variables. The highest removal (95%) was achieved at pH 7.5. It could be concluded that alkaline range pH enables the maximum removal of the organic compound, indicated by the decolorization. It is worth mentioning that a pH level less than 5.0 might cause slower reaction stemming from iron species complexation as well as oxonium ion $[H_3O_2]^+$ formation [35]. Moreover, pH lower than 7.5 could induce the precipitation of iron ions (especially Fe³⁺), that consequently inhibit the regeneration of ferrous ions. Further increase in pH level (>7.5) might cause instability of the hydrogen peroxide to be decomposed into neutral molecules O₂ and H₂O. Herein, the investigation results suggest that the Fenton system was sensitive to pH level and the system should be maintained in alkaline condition to promote iron solubility.



Figure 9. Role of Fe²⁺ concentration and pH in leachate decolorization depicted in 3D response surfaces and contour plot.

Figure 10 exhibits the role of pH in the leachate decolorization, where the data were obtained from a batch experiment using pH values ranging from 3 to 9. According to the investigation, the pH level that could yield the optimum treatment result is pH 7.5. It appears that that alkaline range pH is preferred to produce a maximum amount of $^{\circ}$ OH radicals. When pH solution is set above 7.5, the $^{\circ}$ OH radicals are slower as a result of ferric–hydroxo complex formation [38,41]. The correlation of the H₂O₂:Fe²⁺ ratio and the color removal efficiency were observed positive until the ratio reached 4. However, the correlation became negative once the ratio was increased which is attributed to the inhibition of the scavenging activities of OH $^{\circ}$ radical by excessive amounts of hydrogen peroxide.



$$OH^{\bullet} + H_2O_2 \to HOO^{\bullet} + H_2O \tag{10}$$

Figure 10. Role of pH and H_2O_2 :Fe²⁺ ratio in leachate decolorization depicted in 3D response surfaces and contour plot.

4. Validation of the Experimental Model at Optimized Conditions

Numerical optimization was carried out to obtain the optimum parameters for the color removal from the leachate by using response surface and desirability functions. In this regard, all variables were prepared to fit the designed range, and color removal was set at maximal value. The highest desirability was then revealed as pH = 8.36, H_2O_2 :Fe²⁺ ratio = 3.32, Fe²⁺ concentration = 964.95 mg/L, and reaction time = 50.15 min, in which the color removal was expected to reach 100%. To validate the prediction of the model and the reliability of the optimum combination, an additional batch experiment was conducted using suggested conditions. As a result, 97.68% color removal was obtained. The small error between the experimental and forecast values suggests high reliability of the model. Furthermore, this outcome was better than that of the reported combined Fenton methods using manual optimization [17–19]. Taken together, our study suggests that the RSM is a robust tool for optimizing the working conditions of Fenton-process-based leachate decolorization.

5. Conclusions

To date, landfilling has been the most common method of municipal waste management, due to both technological and economic reasons. A problem associated with the use of landfills is the production of leachate which can lead to serious environmental problems. Characteristics of leachate depend on the composition and amount of deposited waste, the construction and operational conditions of the landfill, and the age of the landfill. Leachate contains high concentrations of ammonia and organic contaminants, halogenated hydrocarbons, heavy metals, and inorganic salts. The Fenton process is used to oxidize complex organic constituents found in wastewater that is difficult to degrade biologically into simpler end products and facilitate the conversion of pollutants to less harmful and more biodegradable compounds. It removes a wide range of organic matter and has been studied by several researchers for the treatment of various types of industrial and residential wastewaters, including landfill leachate. In this regard, mathematical models such as the response surface method (RSM) have proved to be useful for process design and optimization of responses. This model was also used to reduce the number of experimental trials. The RSM provides information and is an economical approach to monitoring the effect of the independent variables on the response. It can reduce the number of experiments that need to be carried out as well as reducing time and expense.

The RSM with CCD could be utilized to optimize the operating conditions of the Fenton process in landfill leachate treatment, where data produced using the RSM has been shown to be statistically reliable. The optimum conditions were as follows: pH = 8.36, H_2O_2 :Fe²⁺ ratio = 3.32, Fe²⁺ concentration = 964.95 mg/L, and reaction time = 50.15 min. The quadratic models are in line with the experimental data (R² > 0.9026). Maximum color removal achieved under the determined optimal conditions was 97.68%, higher than that of previously reported work.

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References

- 1. Maslahati Roudi, A.; Chelliapan, S.; Wan Mohtar, W.H.M.; Kamyab, H. Prediction and optimization of the fenton process for the treatment of landfill leachate using an artificial neural network. *Water* **2018**, *10*, 595. [CrossRef]
- 2. Moravia, W.G.; Amaral, M.C.; Lange, L.C. Evaluation of landfill leachate treatment by advanced oxidative process by Fenton's reagent combined with membrane separation system. *Waste Manag.* **2013**, *33*, 89–101. [CrossRef] [PubMed]
- Anita, M.R.; Sabariah, B.; Shreeshivadasan, C.; Roudi, A.; Hesam, K.; Mohammad, S. Effect of temperature on theperformance of Porous Membrane Activated Sludge Reactor (PMASR) treating synthetic wastewater. *Res. J. Pharm. Biol. Chem. Sci.* 2014, 5, 84–100.
- 4. Dolatabadi, M.; Świergosz, T.; Ahmadzadeh, S. Electro-Fenton approach in oxidative degradation of dimethyl phthalate-The treatment of aqueous leachate from landfills. *Sci. Total. Environ.* **2021**, 772, 145323. [CrossRef]
- Dos Santos Vaz, C.A.; Samanamud, G.L.; da Silva, R.S.; França, A.B.; Quintão, C.M.F.; Urzedo, A.P.; Silva, M.B.; Neto, J.C.B.; Amaral, M.S.; Loures, C.C.A.J. Modeling and optimization of hybrid leachate treatment processes and scale-up of the process. *J. Clean. Prod.* 2021, 312, 127732. [CrossRef]
- 6. Roudi, A.M.; Akhlaghi, E.; Chelliapan, S.; Kaboli, A.; Roudi, A.M.; Aslani, H.; Selvam, S.B. Treatment of Landfill Leachate via Advanced Oxidation Process (AOPs)—A Review. *Res. J. Pharm. Biol. Chem. Sci.* **2015**, *6*, 260–271.
- Marlina; Iqhrammullah, M.; Saleha, S.; Fathurrahmi; Maulina, F.P.; Idroes, R. Polyurethane film prepared from ball-milled algal polyol particle and activated carbon filler for NH3-N removal. *Heliyon* 2020, 6, e04590. [CrossRef]
- 8. El Mrabet, I.; Ihssane, B.; Valdés, H.; Zaitan, H. Optimization of Fenton process operating conditions for the treatment of the landfill leachate of Fez city (Morocco). *Int. J. Environ. Sci. Technol.* **2021**, *6*, 1–14. [CrossRef]

- 9. Zolfaghari, M.; Jardak, K.; Drogui, P.; Brar, S.K.; Buelna, G.; Dubé, R. Landfill leachate treatment by sequential membrane bioreactor and electro-oxidation processes. *J. Environ. Manag.* **2016**, *184*, 318–326. [CrossRef]
- 10. Erkan, H.S.; Apaydin, O. Final treatment of young, middle-aged, and stabilized leachates by Fenton process: Optimization by response surface methodology. *Desalination Water Treat*. **2015**, *54*, 342–357. [CrossRef]
- 11. Kumar, S.S.; Malyan, S.; Kumar, A.; Bishnoi, N.R. Optimization of Fenton's Oxidation by Box-Behnken Design of Response Surface Methodology for Landfill Leachate. J. Mater. Environ. Sci. 2016, 7, 4456–4466.
- 12. Guvenc, S.Y.; Varank, G. Degradation of refractory organics in concentrated leachate by the Fenton process: Central composite design for process optimization. *Front. Environ. Sci. Eng.* **2021**, *15*, 1–16.
- 13. Kermani, M.; Shahsavani, A.; Ghaderi, P.; Kasaee, P.; Mehralipour, J. Optimization of UV-Electroproxone procedure for treatment of landfill leachate: The study of energy consumption. *J. Environ. Health Sci. Eng.* **2021**, *19*, 81–93. [CrossRef]
- 14. Kwarciak-Kozłowska, A.; Fijałkowski, K.L. Efficiency assessment of municipal landfill leachate treatment during advanced oxidation process (AOP) with biochar adsorption (BC). *J. Environ. Manag.* **2021**, *287*, 112309. [CrossRef]
- 15. Mahtab, M.S.; Islam, D.T.; Farooqi, I.H. Optimization of the process variables for landfill leachate treatment using Fenton based advanced oxidation technique. *Eng. Sci. Technol. Int. J.* 2021, 24, 428–435. [CrossRef]
- 16. Muszyńska, J.; Bąk, Ł.; Górski, J.; Górska, K.; Sałata, A.; Gawdzik, J. Fenton Process Optimization with Landfill Leachate in Janczyce as an Example. *Pol. J. Environ. Stud.* **2021**, *30*, 3769–3775. [CrossRef]
- 17. Silveira, J.E.; Zazo, J.A.; Pliego, G.; Casas, J.A. Landfill leachate treatment by sequential combination of activated persulfate and Fenton oxidation. *Waste Manag.* **2018**, *81*, 220–225. [CrossRef]
- Yazici Guvenc, S.; Dincer, K.; Varank, G. Performance of electrocoagulation and electro-Fenton processes for treatment of nanofiltration concentrate of biologically stabilized landfill leachate. J. Water Process. Eng. 2019, 31, 100863. [CrossRef]
- 19. Tejera, J.; Miranda, R.; Hermosilla, D.; Urra, I.; Negro, C.; Blanco, Á. Treatment of a Mature Landfill Leachate: Comparison between Homogeneous and Heterogeneous Photo-Fenton with Different Pretreatments. *Water* **2019**, *11*, 1849. [CrossRef]
- Taşcı, S.; Özgüven, A.; Yıldız, B. Multi-Response/Multi-Step Optimization of Heterogeneous Fenton Process with Fe₃O₄ Catalyst for the Treatment of Landfill Leachate. *Water Air Soil Pollut.* 2021, 232, 275. [CrossRef]
- 21. Kirmizakis, P.; Tsamoutsoglou, C.; Kayan, B.; Kalderis, D. Subcritical water treatment of landfill leachate: Application of response surface methodology. *J. Environ. Manag.* 2014, 146, 9–15. [CrossRef]
- 22. Aslani, H.; Nabizadeh, R.; Nasseri, S.; Mesdaghinia, A.; Alimohammadi, M.; Mahvi, A.H.; Rastkari, N.; Nazmara, S. Application of response surface methodology for modeling and optimization of trichloroacetic acid and turbidity removal using potassium ferrate (VI). *Desalination Water Treat*. **2016**, *57*, 25317–25328. [CrossRef]
- 23. Roudi, A.M.; Chelliapan, S.; Armaki, S.S.M. Determination of Cod and Color Reduction of Stabilized Landfill Leachate by Fenton Process. J. Asian Sci. Res. 2017, 7, 77. [CrossRef]
- 24. Lak, M.G.; Sabour, M.R.; Amiri, A.; Rabbani, O. Application of quadratic regression model for Fenton treatment of municipal landfill leachate. *Waste Manag.* 2012, *32*, 1895–1902.
- 25. Sabour, M.R.; Lak, M.G.; Rabbani, O. Evaluation of the main parameters affecting the Fenton oxidation process in municipal landfill leachate treatment. *Waste Manag. Res.* 2011, 29, 397–405. [CrossRef]
- Wu, Y.; Zhou, S.; Qin, F.; Ye, X.; Zheng, K. Modeling physical and oxidative removal properties of Fenton process for treatment of landfill leachate using response surface methodology (RSM). J. Hazard. Mater. 2010, 180, 456–465. [CrossRef]
- Amiri, A.; Sabour, M.R. Multi-response optimization of Fenton process for applicability assessment in landfill leachate treatment. Waste Manag. 2014, 34, 2528–2536. [CrossRef]
- Santin-Gusman, M.; Moreno-Andrés, J.; Cisneros-Abad, M.; Aguilar-Ramírez, S. Optimization for Fenton Process in Removal of COD for Landfill Leachate Treatment. Int. J. Environ. Sci. Dev. 2015, 6, 920–924.
- 29. Bianco, B.; De Michelis, I.; Vegliò, F. Fenton treatment of complex industrial wastewater: Optimization of process conditions by surface response method. *J. Hazard. Mater.* **2011**, *186*, 1733–1738. [CrossRef]
- Myers, R.; Montgomery, D.; Anderson-Cook, C. Response Surface Methodology: Process and Product Optimization Using Designed Experiments, 3rd ed.; John Wiley: Hoboken, NJ, USA, 2009.
- 31. Montgomery, D.C. Design and Analysis of Experiments; John Wiley: Hoboken, NJ, USA, 2009.
- 32. Ghafari, S.; Aziz, H.A.; Isa, M.H.; Zinatizadeh, A.A. Application of response surface methodology (RSM) to optimize coagulation– flocculation treatment of leachate using poly-aluminum chloride (PAC) and alum. J. Hazard. Mater. 2009, 163, 650–656. [CrossRef]
- 33. Talebi, A.; Ismail, N.; Teng, T.T.; Alkarkhi, A.F. Optimization of COD, apparent color, and turbidity reductions of landfill leachate by Fenton reagent. *Desalination Water Treat.* **2014**, *52*, 1524–1530. [CrossRef]
- Roudi, A.M.; Kamyab, H.; Chelliapan, S.; Ashokkumar, V.; Kumar, A.; Yadav, K.K.; Gupta, N. Application of response surface method for Total organic carbon reduction in leachate treatment using Fenton process. *Environ. Technol. Innov.* 2020, 19, 101009. [CrossRef]
- Kwon, B.G.; Lee, D.S.; Kang, N.; Yoon, J. Characteristics of p-chlorophenol oxidation by Fenton's reagent. Water Res. 1999, 33, 2110–2118. [CrossRef]
- Ahmadian, M.; Reshadat, S.; Yousefi, N.; Mirhossieni, S.H.; Zare, M.R.; Ghasemi, S.R.; Rajabi Gilan, N.; Khamutian, R.; Fatehizadeh, A. Municipal leachate treatment by Fenton process: Effect of some variable and kinetics. *J. Environ. Public Health* 2013, 2013, 169682. [CrossRef]

- 37. Tengrui, L.; Al-Harbawi, A.; Jun, Z.; Bo, L.M. The effect and its influence factors of the Fenton process on the old landfill leachate. *J. Appl. Sci.* 2007, *7*, 724–727. [CrossRef]
- 38. Mohajeri, S.; Aziz, H.A.; Isa, M.H.; Bashir, M.J.; Mohajeri, L.; Adlan, M.N. Influence of Fenton reagent oxidation on mineralization and decolorization of municipal landfill leachate. *J. Environ. Sci. Health Part A* **2010**, *45*, 692–698. [CrossRef]
- 39. Lopez, A.; Pagano, M.; Volpe, A.; Di Pinto, A.C. Fenton's pre-treatment of mature landfill leachate. *Chemosphere* 2004, 54, 1005–1010. [CrossRef]
- 40. Zhang, H.; Choi, H.J.; Huang, C.P. Optimization of Fenton process for the treatment of landfill leachate. *J. Hazard. Mater.* 2005, 125, 166–174. [CrossRef]
- 41. Zhang, H.; Fei, C.; Zhang, D.; Tang, F. Degradation of 4-nitrophenol in aqueous medium by electro-Fenton method. *J. Hazard. Mater.* **2007**, 145, 227–232. [CrossRef]