

Modeling and Multi-objective Optimization of a Packed Bed Reactor for Sulfur Dioxide Removal by Magnesium Oxide Using Non-dominated Sorting Genetic Algorithm II



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Nowadays, protecting the environment is of utmost importance worldwide, and sulfur dioxide is one of the main pollutants in the atmosphere. This work proposes a new method for simultaneous SO₂ removal by MgO, and production of magnesium sulfate in a packed bed reactor for which breakthrough curves have been obtained. Furthermore, the effect of important operating parameters, including temperature, SO₂ concentration, and gaseous flow rate was investigated. Experiments showed that increasing the temperature improved the breakthrough lifetime, but the increase in concentration and flow rate reduced the lifetime. The experimental results were predicted successfully by applying the Random Pore Model (RPM). Finally, the Non-dominated Sorting Genetic Algorithm II (NSGA II) that is a technique for multi-objective optimization, was employed to determine the best operating parameters for SO₂ removal by magnesium oxide in the packed bed reactor.

Keywords

SO₂ removal, magnesium oxide, packed bed reactor, random pore model, modeling and simulation, multi-objective optimization

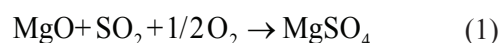
Introduction

Nowadays, protecting the environment, and air pollutant removal processes are of great importance all over the world. Acid rain is one of the major environmental problems, and it can destroy the forests and aquatic animals. The most important factor resulting in the formation of acid rains is the presence of SO₂ in the atmosphere. Sulfuric acid is produced as the result of SO₂ and water vapor reaction, and falling to the ground with rain^{1,2}.

The main sources of SO₂ emission are metallurgical and coal-fired power plants (stationary units), and vehicles that consume high sulfur content fuel (mobile units)³. Many attempts have been made in order to reduce the sulfur dioxide emission to the atmosphere in the world. In many countries, the sulfur content of gasoline and gas oil should be below 10 ppm (Euro 5 standard)⁴. These low-sulfur-content fuels are produced by deep hydrotreating processes in refineries^{5,6}. Furthermore, in stationary units, flue gas desulfurization (FGD) processes are applied to adsorb or absorb sulfur dioxide and control the level of SO₂ in the exhaust gases.

FGD processes are categorized into the throw-away systems for low SO₂ concentration, including coal-fired power plants, and regenerative systems for high SO₂ concentration such as metallurgical units³. The common adsorbents in throw-away systems are CaO, MgO, and Fe₂O₃.

The main reaction in throw-away FGD systems based on MgO is:



On the other hand, magnesium sulfate is usually produced as a result of reaction between MgO and sulfuric acid. MgSO₄ is mainly used in agriculture as a fertilizer. The increasing global demand for agricultural products has resulted in the increase of MgSO₄ production. The other applications of magnesium sulfate are in medical, food preparation, and construction industries⁷.

The conventional methods of magnesium sulfate production are extraction of mineral kieserite, and MgO reaction with H₂SO₄. In the first method, kieserite is dissolved in water and MgSO₄ product is obtained by crystallization process. In the second method, MgO (from magnesite calcination) is reacted with H₂SO₄ producing MgSO₄ and then crystallization is used to increase the purity of the final product⁸.

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The conducted research on the ability of MgO to capture SO₂ in FGD systems are much fewer, compared to the calcium oxide. Jae *et al.* studied MgO-based sorbents in a fixed bed reactor to adsorb sulfur dioxide⁹. They prepared regenerable sorbents by adding cerium and iron as additives into the MgO-based sorbents. Based on their results, the efficiency of the sorbents and the rate of reaction had improved by using these additives.

Zhang *et al.* adsorbed SO₂ by using natural magnesite¹⁰. They declared that calcination temperature is a key parameter and that it affects adsorption capacity of sorbents. The adsorption capacity was reported to be 140.7 mg g⁻¹.

Prezepiorski *et al.* studied the adsorption of SO₂ from air by using MgO/carbon sorbents¹¹. Furthermore, they considered the effect of operating parameters such as temperature, porosity, and humidity, and analyzed the adsorption mechanism.

Zermeno *et al.* used natural magnesite in packed bed reactor and obtained the experimental breakthrough profiles of the reactor¹². They announced that natural magnesite had a good adsorption capacity. They also declared that natural magnesite could be considered as a good alternative for CaO.

Lee *et al.* studied the removal of sulfur dioxide by magnesium oxide¹³. They improved the adsorption capacity of sorbents by adding titanium dioxide. Their sorbents were prepared by co-precipitation method. They improved the adsorption capacity of the sorbent by adding TiO₂ from 38 g sulfur g⁻¹ to 44 g sulfur g⁻¹.

Based on the analysis carried out by Magnabosco, MgO can be used as a suitable adsorbent for SO₂ removal in the regenerator of FCC unit¹⁴.

Li *et al.* found that the operating cost of FGD process based on MgO is lower than the FGD process based on CaO or CaCO₃¹⁵.

Yi *et al.* evaluated MgO for simultaneous adsorption of SO₂ and NO_x from coal-fired power plant flue gas¹⁶. The performance of different morphologies of MgO was also considered in the study.

Liato *et al.* examined MgO in the wet FGD process for SO₂ removal¹⁷. They announced that MgO showed a high capacity for adsorbing SO₂.

Zou *et al.* used MgO for simultaneous removal of SO₂ and NO_x with oxidation of ozone¹⁸. Furthermore, the effect of O₃/NO molar ratio, oxidation temperature and other operating parameters were considered. They reported that increasing SO₂ concentration had a negative effect on NO_x removal efficiency.

Wang *et al.* studied oxidation of SO₂ to sulfate on surface of MgO by diffuse reflectance infrared Fourier transform spectroscopy¹⁹. They revealed

that the product of MgO+SO₂ reaction was sulfite, bisulfite, and sulfate.

Zhao and Zou tested micro-sized MgO slurry and MgO nanofluids slurry in wet desulfurization process²⁰. They found that the efficiency of MgO nanofluids was better than that of micro-sized MgO slurry.

Many attempts have been made to model gas-solid reactions. We were able to predict the performance of real systems by efficient application of modeling and simulation²¹. Furthermore, modeling and simulation along with the application of mathematical tools such as genetic algorithm, can lead to the discovery of optimum operating parameters of the various processes.

Dry FGD process is an important type of non-catalytic gas-solid reactions that SO₂ adsorbed by adsorbents like MgO, CaO, and CuO. Kinetic study of these reactions is an appropriate tool for improvement of FGD process through design and optimization of the related reactors. For this reason, various models including shrinking core model, volume reaction model, grain model, modified grain model, nucleation model, single pore model, and random pore model (RPM) are developed. RPM and modified grain model are able to consider incomplete conversion due to volume increase of the pellet during the reaction. Complicated RPM theory assumes that the reaction occurs on the inner surfaces of cylindrical pores of the sorbent in the series of holes that have a pore size distribution (PSD).

Bahrami *et al.* used RPM to simulate SO₂ removal by CuO as a regenerative process²². The experimental operating parameters in their work included temperature (400–600 °C) and sulfur dioxide concentration (1250–5000 ppm). They compared the experimental data with the values predicted by RPM. They declared that RPM could predict the experimental conversion-time data with a high degree of accuracy.

Moshiri *et al.* adsorbed SO₂ in a packed bed reactor by using CaO²³. Their experimental breakthrough curves were modeled by RPM. They found that RPM could accurately predict the experimental data.

Betancur *et al.* studied gasification process by thermogravimetry and modeled the experimental data by grain model, RPM and hybrid modification RPM²⁴. They concluded that RPM and hybrid modified RPM were the most precise models.

Lopez *et al.* considered CaO reaction with carbon dioxide, and evaluated the effect of inert support²⁵. They applied RPM to calculate the kinetic parameters.

Montagnaro *et al.* modeled the CaO+SO₂ reaction by RPM²⁶. They then compared the experimen-

tal results with the RPM. They announced that the model could correlate the experimental data.

Removal of sulfur dioxide with CuO in a thermogravimeter was examined by Yu *et al.*²⁷ The experimental results were predicted by volume reaction model, grain size model, RPM and pore-blocking model.

Nouri *et al.* evaluated the adsorption of CO₂ by lime in a packed bed reactor and obtained the experimental breakthrough curves²⁸. Furthermore, the acid washing technique was used in their study to improve the adsorption capacity of lime. Finally, they predicted the experimental breakthrough curves with RPM.

In our previous studies, the kinetic parameters of MgO+SO₂ reaction were calculated by RPM and the adsorption capacity of MgO was improved by acid washing method, and the experimental conversion-time and breakthrough curves were compared with the natural sample^{29,30}. Finally, RPM was applied to model the experimental data.

Process optimization has become an interesting area of research during recent years, and as a result, attempts have been made to find the best operating conditions for different chemical processes. Optimization will maximize the process performance and minimize the operating costs. Furthermore, the optimization results can be used as the optimum conditions for large-scale industrial units operations^{31,32}.

Wu *et al.* applied multi-objective optimization for hydrotreating process³³. They optimized the process to minimize the operating cost and SO₂ emission to the environment. They found that the operating cost and emission of sulfur dioxide would be reduced in high temperatures and low pressures.

Zhou *et al.* considered multi-objective optimization of SO₂ removal by activated coke and found the best sorbent preparation parameters³⁴. They found that the best operating conditions were at 924 °C, oxygen concentration 5.9 %, and vapor concentration 20 %.

Bayon *et al.* optimized the operation of a thermal power plant to minimize SO₂ and NO_x emissions to the environment³⁵.

Bakhshi Ani *et al.* simulated a trickle-bed hydrotreating reactor and used multi-objective optimization to find the best operating conditions of the process⁵. They also considered the effect of operating parameters such as temperature, pressure, LHSV, and H₂/oil.

Liu *et al.* simulated desulfurization tower by CFD tool and optimized the related operating parameters³⁶. They found that CFD could accurately predict the experimental data.

To the best of our knowledge, no previous research has been conducted on RPM for MgO+SO₂

packed bed reaction, and a few attempts have been made to investigate the behavior of packed bed reactor for adsorption of SO₂ by MgO. Furthermore, no one has reported the best operating conditions for SO₂ removal in packed bed reactor so far. Therefore, in the present study, the tests were performed to obtain the experimental breakthrough curves of natural MgO in the packed bed reactor. The effects of important operating parameters, such as SO₂ concentration, temperature, and gaseous flow rate on lifetime of the breakthrough curves were investigated experimentally. Then, RPM as a comprehensive and precise model that considers the structural changes of sorbents was applied to simulate the experimental results. The possibility of magnesium sulfate production as a useful by product in the SO₂ removal process by MgO is also considered.

Finally, NSGA II as a technique for multi-objective optimization was employed to determine the best operating parameters for SO₂ removal by magnesium oxide in the packed bed reactor.

Materials and methods

Natural magnesium carbonate/oxide was supplied from the Nehbandan mine (Iran), in the form of spherical particles of about 6 mm in diameter. The composition of the mineral sample is presented in Table 1. The data in this table was obtained from the XRF analysis results. Highly pure SO₂ (99.95 %) and zero air (mixture of pure oxygen and nitrogen) were the gases used in this research.

The experimental setup included tubular reactor (310 stainless steel, 1.5 cm inner diameter, and 6 cm height), K-type thermocouple, vertical furnace, two MFCs (mass flow controller), and online mass spectrometer (MS). The schematic diagram of the apparatus is presented in Fig. 1.

The calcination process was applied to prepare a highly porous MgO from magnesium carbonate/oxide directly in the packed bed reactor as part of the experimental setup. In this process, the sample was heated to the operating reaction temperature (500–600 °C) for about 30 min under zero-air stream. The porosity of the sample increased due to the release of H₂O and CO₂ molecules from mineral magnesium carbonate/oxide.

After the calcination process, SO₂ was mixed in zero-air in the predefined concentration, and it was injected to the reactor to start the adsorption/reaction. MS continuously monitored outlet gases from the reactor. The experimental breakthrough curves were obtained from MS results. Table 2 shows the operating conditions of the experiments.

Table 1 – XRF analysis of the natural magnesium carbonate/oxide from Nehbandan mine

Component	wt%
Na ₂ O	0.04
MgO	92.94
SiO ₂	3.52
P ₂ O ₅	0.03
SO ₃	0.09
Cl	0.05
CaO	1.64
MnO	0.23
Fe ₂ O ₃	1.07
NiO	0.36

Table 2 – Operating conditions of the packed bed reactor

Property	value
Packed bed length (cm)	6
Zero-air flow (m ³ s ⁻¹)	1.67·10 ⁻⁶ – 3.34·10 ⁻⁶
SO ₂ concentration (vol.%)	1–2
Temperature (°C)	500–600

Packed bed reactor modeling

RPM as a comprehensive and real model was employed in this study to simulate the packed bed reactor behavior and predict the breakthrough curves. This model was developed by Bhatia and Perlmutter and its main assumptions are^{37–39}:

- Pseudo-steady state approximation;
- Negligible bulk flow effect;
- Isothermal condition for system;
- Irreversible and first order reaction.

The RPM governing equations with the initial and boundary conditions are²²:

$$\frac{1}{y^2} \frac{\partial}{\partial y} (\delta y^2 \frac{\partial a}{\partial y}) = \frac{\phi^2 ab \sqrt{1-\psi \ln b}}{1 + \frac{\beta Z}{\psi} [\sqrt{1-\psi \ln b} - 1]} \quad (2)$$

$$\frac{\partial b}{\partial \theta} = - \frac{ab \sqrt{1-\psi \ln b}}{1 + \frac{\beta Z}{\psi} [\sqrt{1-\psi \ln b} - 1]} \quad (3)$$

$$\theta = 0: \quad b = 1 \quad (4)$$

$$y = 0: \quad \frac{\partial a}{\partial y} = 0 \quad (5)$$

$$y = 1: \quad \frac{\partial a}{\partial y} = \frac{Sh}{\delta} (1 - a) \quad (6)$$

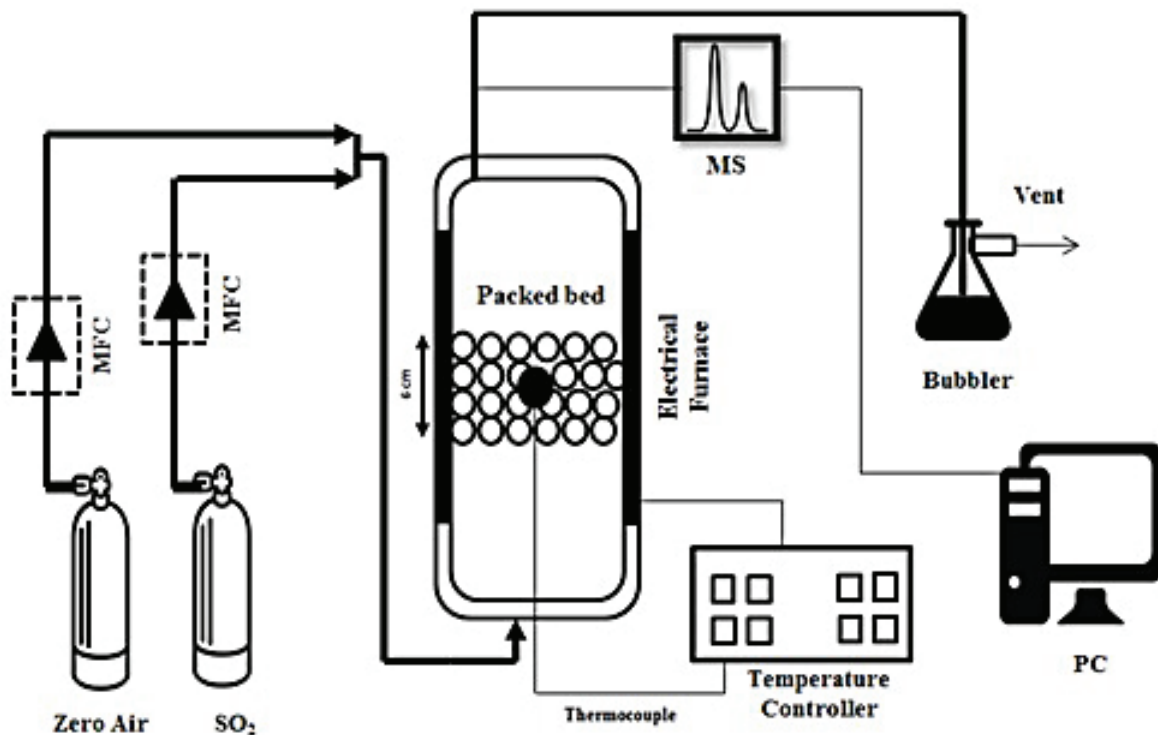


Fig. 1 – Packed bed reactor setup with outlet gas analysis by mass spectrometer

In equation (1), ψ is the main RPM parameter obtained from the whole PSD of the calcined sample. The packed bed reactor differential equations were derived from the mass balance as follows³⁷:

For bulk gas in the reactor:

$$\frac{\partial^2 w}{\partial \zeta^2} - Pe \frac{\partial w}{\partial \zeta} = \alpha(w - \zeta_{y=1}) \quad (7)$$

$$\zeta = 0: \quad w = 1 \quad (8)$$

$$\zeta = \Lambda: \quad \frac{\partial w}{\partial \zeta} = 0 \quad (9)$$

For the reactant gas in the pellet:

$$\frac{1}{y^2} \frac{\partial}{\partial y} \left(y^2 \frac{\partial \zeta}{\partial y} \right) = \phi^2 f(x) \zeta \quad (10)$$

$$y = 0: \quad \frac{\partial \zeta}{\partial y} = 0 \quad (11)$$

The $f(x)$ in equation (10) is inserted from the right-hand side of equation (3).

For the solid reactant:

$$\frac{\partial X}{\partial \tau} = f(x) \zeta \quad (12)$$

$$\tau = 0: \quad X = 0 \quad (13)$$

The variables of the packed bed reactor are expressed as follows:

$$\zeta = \frac{x}{R} \quad (14)$$

$$y = \frac{r_p}{R} \quad (15)$$

$$\zeta = \frac{C_A}{C_{Ab0}} \quad (16)$$

$$w = \frac{C_{Ab}}{C_{Ab0}} \quad (17)$$

$$\tau = \nu_B \frac{kC_{Ab0}t}{C_{B0}} \quad (18)$$

$$\Lambda = \frac{L}{R} \quad (19)$$

The parameters of the system are presented as follows:

$$Pe = \frac{uR}{D_L} \quad (20)$$

$$\alpha = \frac{(1-\varepsilon) 3k_m R}{\varepsilon D_L} \quad (21)$$

$$\phi^2 = \frac{\nu_A k R^2}{D_e} \quad (22)$$

$$Bi_m = \frac{k_m R}{D_e} \quad (23)$$

The energy consumption in calcination process to reach and maintain the reactor at desired temperature can be calculated from the following equation based on the energy balance:

$$Q = m_{ad} C_{p,ad} (T_{cal} - T_0) + m_r C_{p,r} (T_{cal} - T_0) + hA_r (T_{cal} - T_0) t_{cal} \quad (24)$$

The solution method for solving nonlinear RPM and packed bed reactor partial differential equations was developed in MATLAB® based on the finite element method. The details about the solution method has been already described in our previous study²³. The explanation for all of the symbols that have been used in Eq. (2) to (24) are given in the nomenclature part.

Optimization

One of the well-known optimization tools is genetic algorithm (GA) that is based on Darwinian evolution. The main operators of GAs are crossover, recombination, mutation, and selection⁴⁰. The first step in GA is the random selection of a population from the parent chromosomes. The evolution of selected population to the better chromosomes is applied by GA main operators. The chromosomes for recombination are chosen in the selection stage. The new offspring chromosomes are generated in recombination stage from two parents after finishing the selection stage. Then, the children enter the mutation stage. In the mutation stage, new features are added to the population by creating new chromosomes from one child⁴¹.

Srinivas and Deb introduced the Non-dominated Sorting Genetic Algorithm (NSGA) in 1995. In 2000, Deb proposed NSGA II to solve some difficulties of NSGA⁴². NSGA II utilizes the elite strategy to sort parents' and children populations. The diversity of solutions was enhanced by using crowded comparison operator in NSGA II. In this study, the evaluation tool in NSGA II was the packed bed reactor model and the energy consumption equation. The calculation work flow chart of NSGA II is presented in Fig. 2.

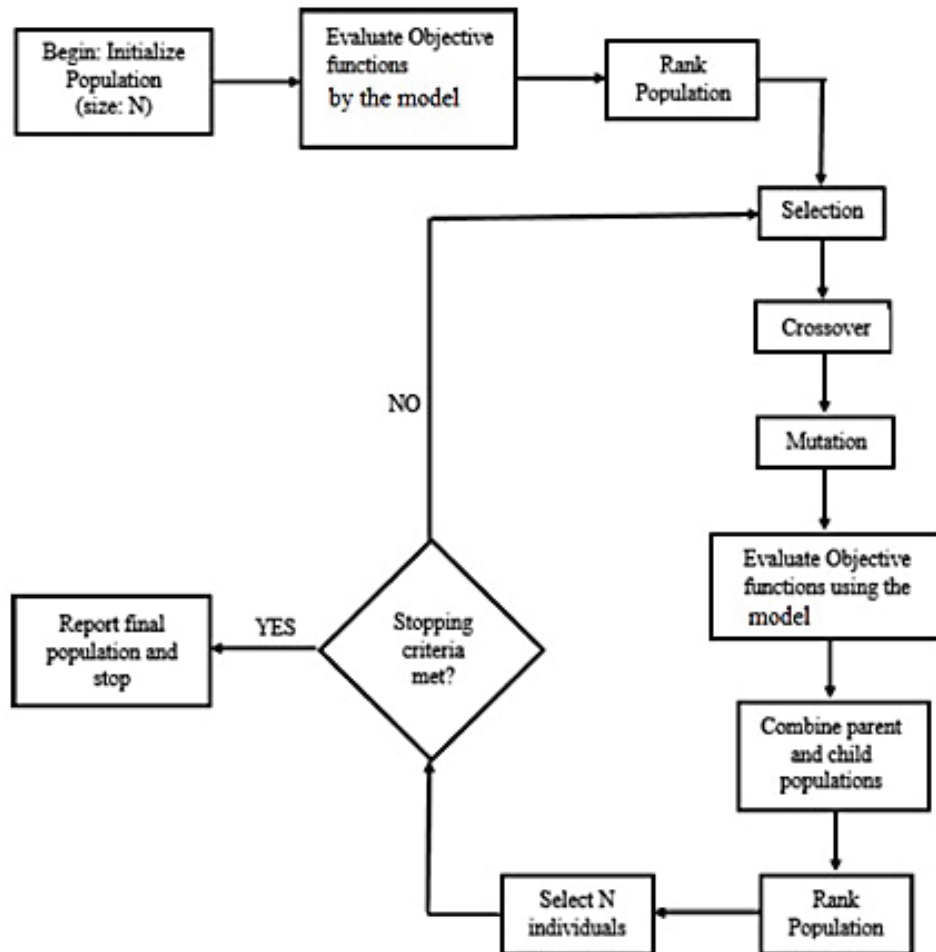


Fig. 2 – Calculation work flow chart of NSGA II

Results and discussion

Model validation

As mentioned, packed bed reactors with suitable sorbents can be used in air pollutants removal processes, such as FGD (SO₂ elimination), and CO₂ concentrating from flue gases. In these reactors, evolution of the breakthrough curve (measuring pollutant outlet concentration versus time) is essential, since breakthrough curve determines effective lifetime of a packed bed reactor in the aforementioned environmental engineering processes. Consequently, mathematical modeling for accurate prediction of breakthrough curve of a packed bed reactor is of great importance.

The experimental breakthrough curve of exit sulfur dioxide from the natural MgO reaction in the packed bed at 600 °C, 1 vol. % SO₂ concentration, and 1.67·10⁻⁶ m³ s⁻¹ (100 cc min⁻¹) zero-air flow (base case) is presented in Fig. 3. Based on the results, the breakthrough time is about 210 min.

The breakthrough curve of natural magnesium oxide in the packed bed was predicted by the RPM.

Table 3 – Required parameters for RPM

Parameter	Value	
	500 °C	600 °C
D_L (m ² s ⁻¹)	$2.2 \cdot 10^{-6}$	$2.7 \cdot 10^{-6}$
\bar{r} (m)	$2.66 \cdot 10^{-7}$	
ϵ_0	0.69	
S_0 (cm ⁻¹)	$6.32 \cdot 10^6$	
L_0 (cm ⁻²)	$6.1 \cdot 10^{12}$	
ψ	0.58	
k_s	$2.38 \cdot 10^{-3} \exp\left(\frac{-38629.33}{RT}\right)$	
D_p	$3.28 \cdot 10^{-14} \exp\left(\frac{-68474.1}{RT}\right)$	

The required parameters to predict the curve by RPM are represented in Table 3 from our previous studies^{29,30}. The RPM main parameter (ψ) was calculated from the whole PSD (micro, meso, and macro) of the calcined magnesium oxide pellets.

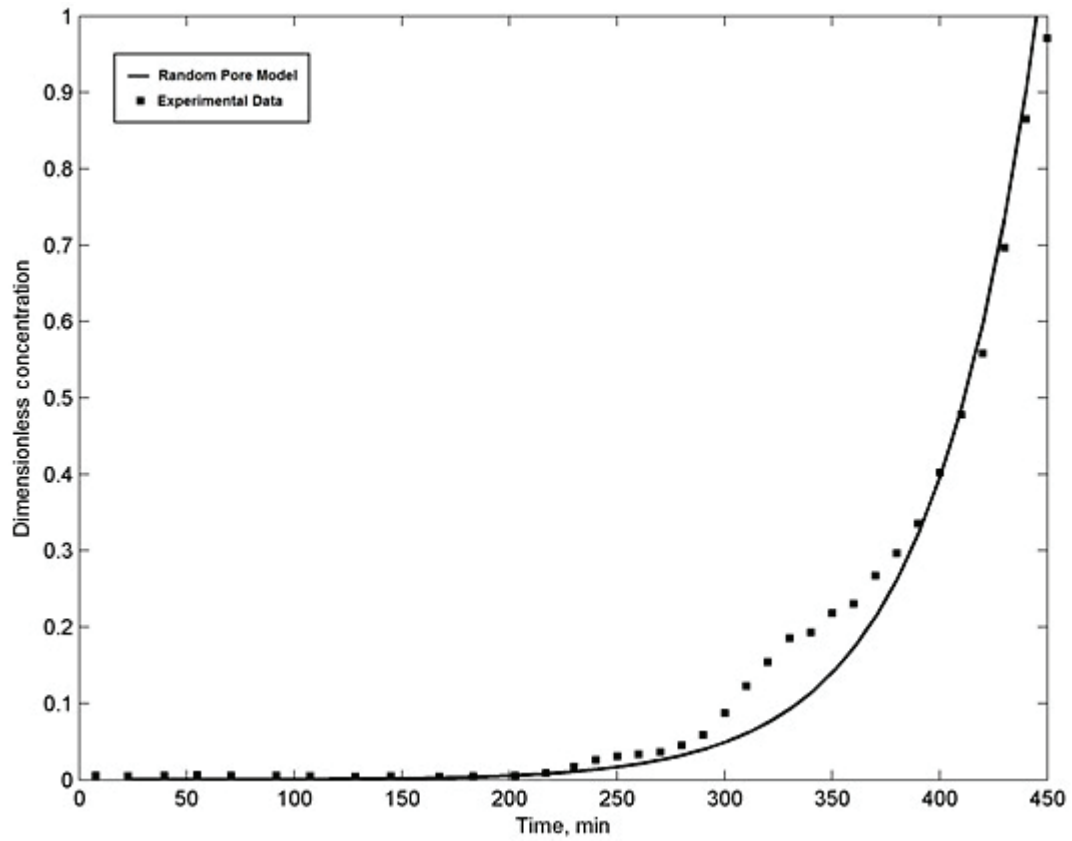


Fig. 3 – Comparison between random pore model predictions and experimental SO_2 breakthrough curves from MgO packed bed reactor at base case ($600\text{ }^\circ\text{C}$, $1\text{ vol.}\% SO_2$ concentration, and $1.67 \cdot 10^{-6}\text{ m}^3\text{ s}^{-1}$ (100 cc min^{-1}) zero-air flow rate)

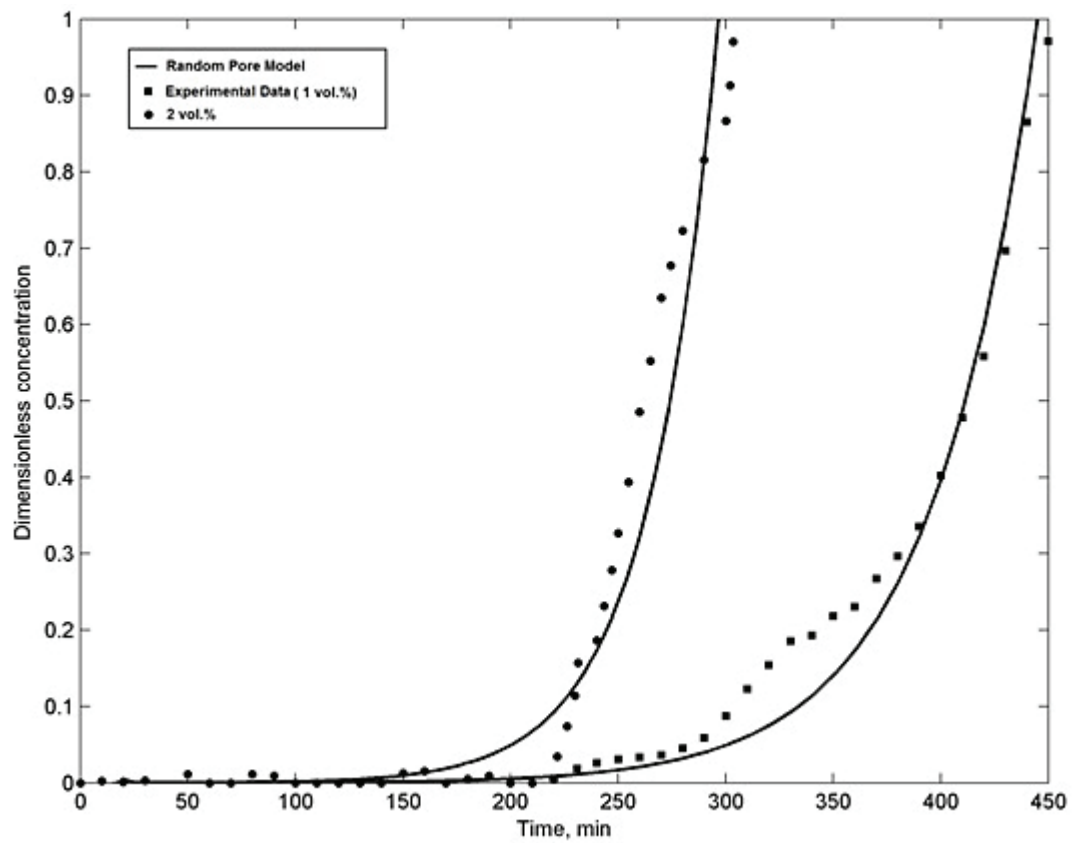


Fig. 4 – Comparison between random pore model predictions and experimental SO_2 breakthrough curves from MgO packed bed reactor at $600\text{ }^\circ\text{C}$, $1.67 \cdot 10^{-6}\text{ m}^3\text{ s}^{-1}$ (100 cc min^{-1}) zero-air flow rate and various SO_2 concentrations

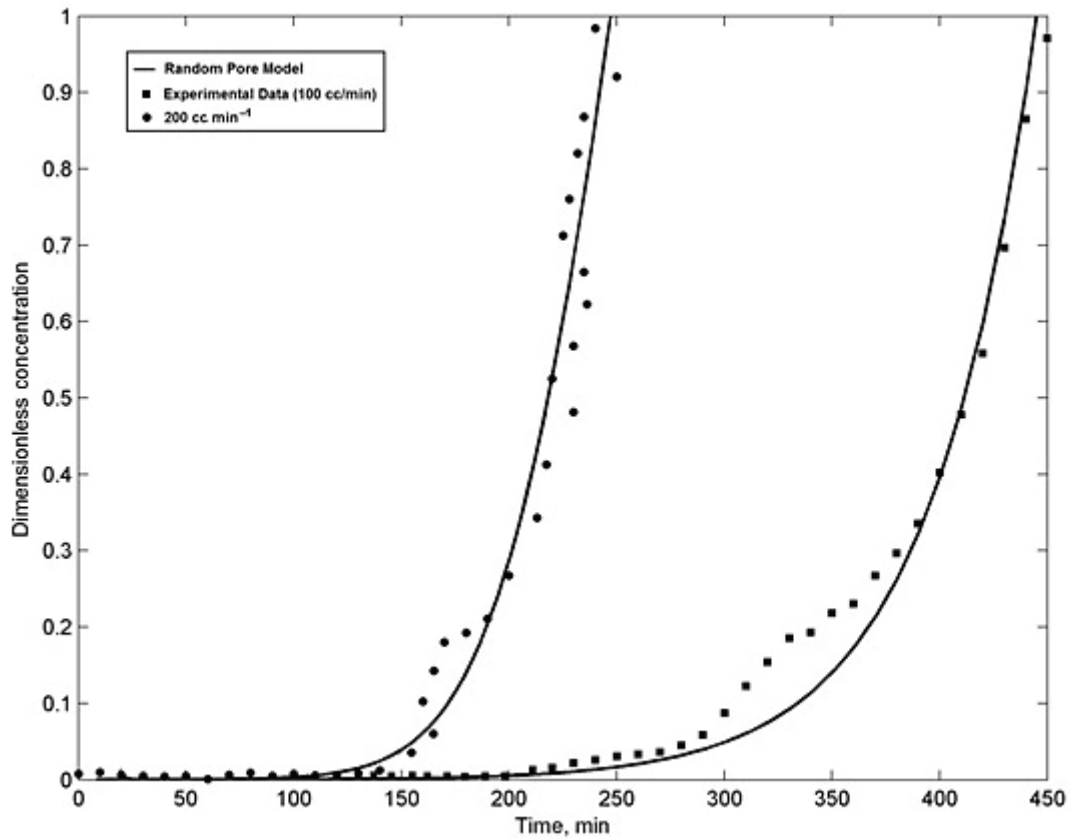


Fig. 5 – Comparison between random pore model predictions and experimental SO_2 breakthrough curves from MgO packed bed reactor at $600\text{ }^\circ\text{C}$, $1\text{ vol.}\%$ SO_2 concentration, and various zero-air flow rates

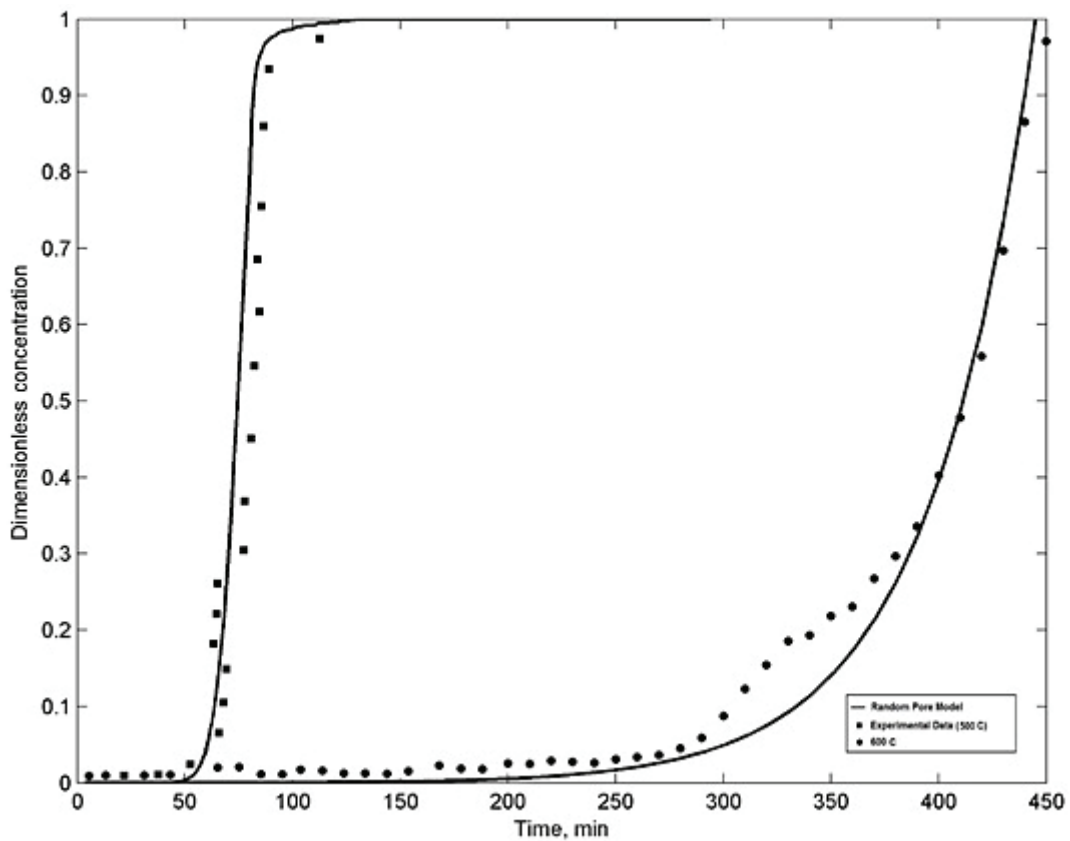


Fig. 6 – Comparison between random pore model predictions and experimental SO_2 breakthrough curves from MgO packed bed reactor at $1\text{ vol.}\%$ SO_2 concentration, $1.67 \cdot 10^{-6}\text{ m}^3\text{ s}^{-1}$ (100 cc min^{-1}) zero-air flow rate and various temperatures

The axial dispersion (D_L) is the sole fitting parameter obtained from the shape of the experimental breakthrough curve. Fig. 3 demonstrates good agreement between the model curve and experimental data.

Fig. 4 illustrates the effect of SO_2 concentration on the breakthrough time. The breakthrough time was reduced from 210 min to 160 min by increasing SO_2 concentration from 1 vol.% (base case) to 2 vol.% due to the input of more sulfur dioxide reactant (versus the base case) to the packed bed reactor.

The effect of zero-air flow rate on the breakthrough time is presented in Fig. 5. The breakthrough time decreased from 210 to 120 min by increasing zero-air flow rate from $1.67 \cdot 10^{-6} \text{ m}^3 \text{ s}^{-1}$ (100 cc min^{-1} , base case) to $3.34 \cdot 10^{-6} \text{ m}^3 \text{ s}^{-1}$ (200 cc min^{-1}). This effect was due to reduction in gaseous flow residence time versus the base case condition⁴³.

The effect of temperature on the breakthrough time of packed bed reactor is shown in Fig. 6. The breakthrough time was reduced from 210 to 50 min by decreasing the temperature from $600 \text{ }^\circ\text{C}$ (base case operating temperature) to $500 \text{ }^\circ\text{C}$. The reduction was the consequence of the decreased reaction rate. Similar behavior was also observed in the adsorption of SO_2 with CuO in the packed bed reactor²².

In Figs. 4 to 6, the values predicted by the model are in relatively good agreement with the experimental breakthrough curve data.

Magnesium sulfate production

During the process of SO_2 removal by MgO, magnesium sulfate is also produced. Magnesium sulfate is widely used in agriculture and various industries. Simultaneous MgSO_4 production as a valuable compound is an advantage of using MgO as adsorbent for SO_2 removal. The combination of SO_2 removal system with MgSO_4 production unit can improve the economic feasibility of FGD unit.

Fig. 7 shows the XRD analysis of the sample, before and after reaction with SO_2 in the packed bed reactor. The analysis shows that MgSO_4 was produced as a result of $\text{MgO} + \text{SO}_2 + 1/2\text{O}_2$ FGD reaction. The average overall conversion of MgSO_4 in the packed bed reactor system from RPM is presented in Fig. 8 for the base case. During the early stages of the process, the kinetics and pore diffusion controlled the reaction. Then, the product layer diffusion controlled the overall reaction, and consequently, the conversion rate decreased. In this process, the produced MgSO_4 is soluble in water and can be easily separated from the unreacted MgO.

Comparison of MgO FGD performance with other sorbents

In this section, the performance of MgO in SO_2 removal process was compared with other common sorbents, such as CuO, CaO, and Fe_2O_3 for sulfur dioxide abatement according to the available literature data in the field. The basis of this comparison were breakthrough time and the rate constant values. It is worth mentioning that both parameters might not be reported in the literature for all of these sorbents.

A complete kinetics study of $\text{SO}_2 + \text{CuO}$ reaction was carried out in a thermogravimeter by Bahrami *et al.*²² They asserted that the rate constant for CuO at $500 \text{ }^\circ\text{C}$ was $4.37 \cdot 10^{-8} \text{ m s}^{-1}$, while this value for MgO was $6.01 \cdot 10^{-6} \text{ m s}^{-1}$ at the same temperature. High reaction rate of MgO with SO_2 showed that it had better adsorption capacity than CuO at an equal residence time.

Bahrami *et al.* also conducted another research on the SO_2 removal in a packed bed reactor by CuO pellets³⁷. They stated that the CuO breakthrough time at $600 \text{ }^\circ\text{C}$ and SO_2 concentration of 2500 ppm was about 550 min. The breakthrough time of MgO curve at $600 \text{ }^\circ\text{C}$ and SO_2 concentration of 10000 ppm was about 200 min. By applying curve fitting on the experimental results of Bahrami *et al.*, at similar operating conditions, we concluded that the breakthrough times for MgO and CuO were approximately the same.

CaO is the common adsorbent in the throw-away FGD systems. The kinetics study of Moshiri *et al.* on $\text{CaO} + \text{SO}_2$ reaction demonstrated that the rate constant of $\text{CaO} + \text{SO}_2$ reaction at $600 \text{ }^\circ\text{C}$ was $3.12 \cdot 10^{-7} \text{ m s}^{-1}$ ⁴⁴. At the same temperature, the reaction rate constant for $\text{MgO} + \text{SO}_2$ was $1.11 \cdot 10^{-5}$, which was higher than the corresponding value of $\text{CaO} + \text{SO}_2$ reaction.

The packed bed tests of $\text{SO}_2 + \text{CaO}$ reaction were carried out by Dasgupta *et al.*⁴⁵ The results of their study indicated that the maximum breakthrough time in CaO was 60 min obtained at SO_2 concentration of 1 vol.%, temperature of $950 \text{ }^\circ\text{C}$ and reactor length of 15 cm. For MgO, the breakthrough time at SO_2 concentration of 1 vol.%, with reactor length of 6 cm, and temperature of $600 \text{ }^\circ\text{C}$ was 210 min. This implied higher adsorption capacity of MgO in comparison with CaO. The better performance of MgO reduced the number of adsorbent replacement cycles of the packed bed reactor, and as a result, the efficiency of the process would increase and operating costs decrease.

Selvakumar investigated the removal of SO_2 by ferric oxide in a packed bed reactor⁴⁶. The results of their study indicated that the breakthrough time of Fe_2O_3 at $500 \text{ }^\circ\text{C}$, SO_2 concentration of $7.86 \cdot 10^{-3}$

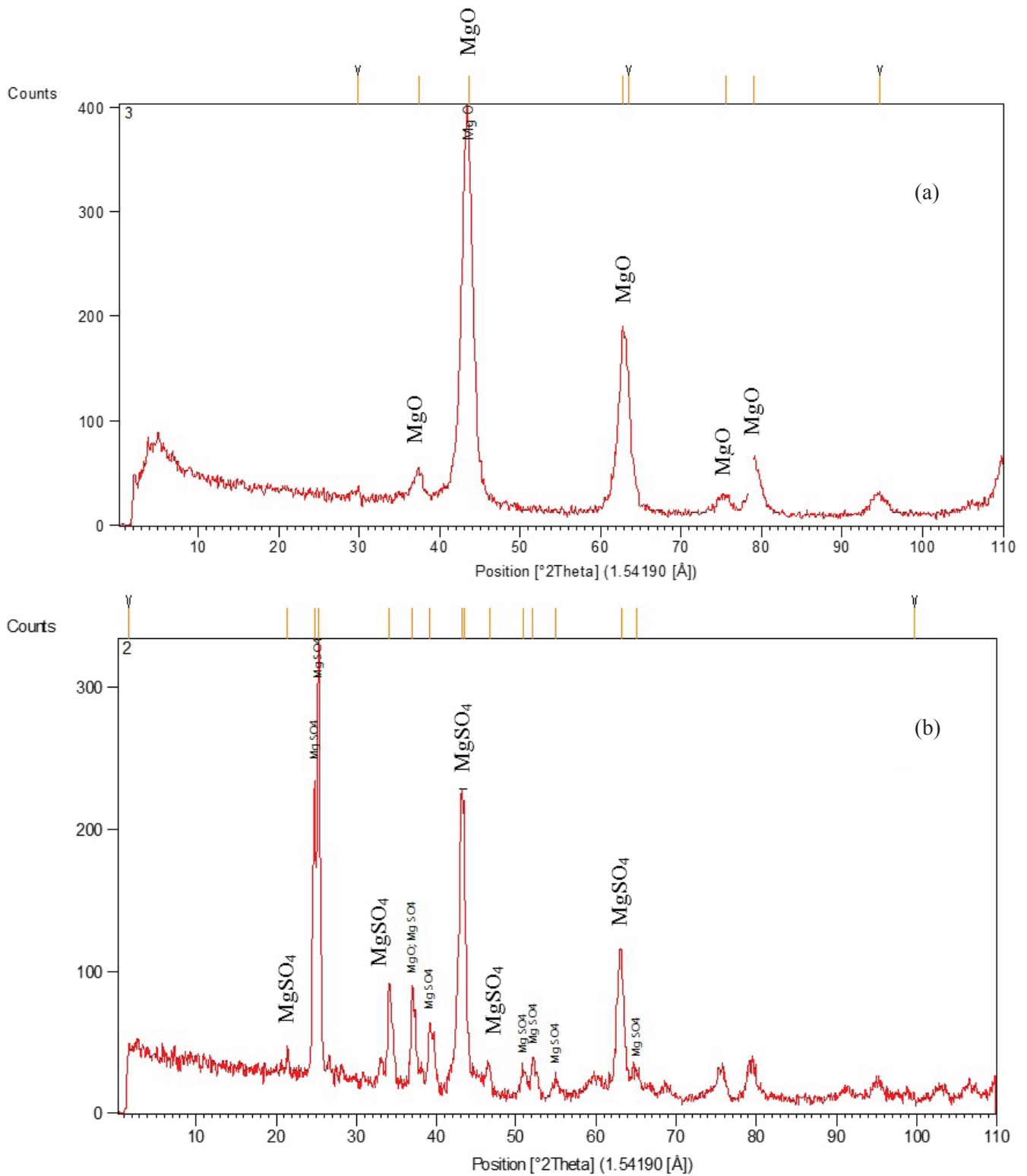


Fig. 7 – XRD analysis of the magnesite sample (a) after calcination process, and (b) after reaction with SO_2 in the packed bed reactor

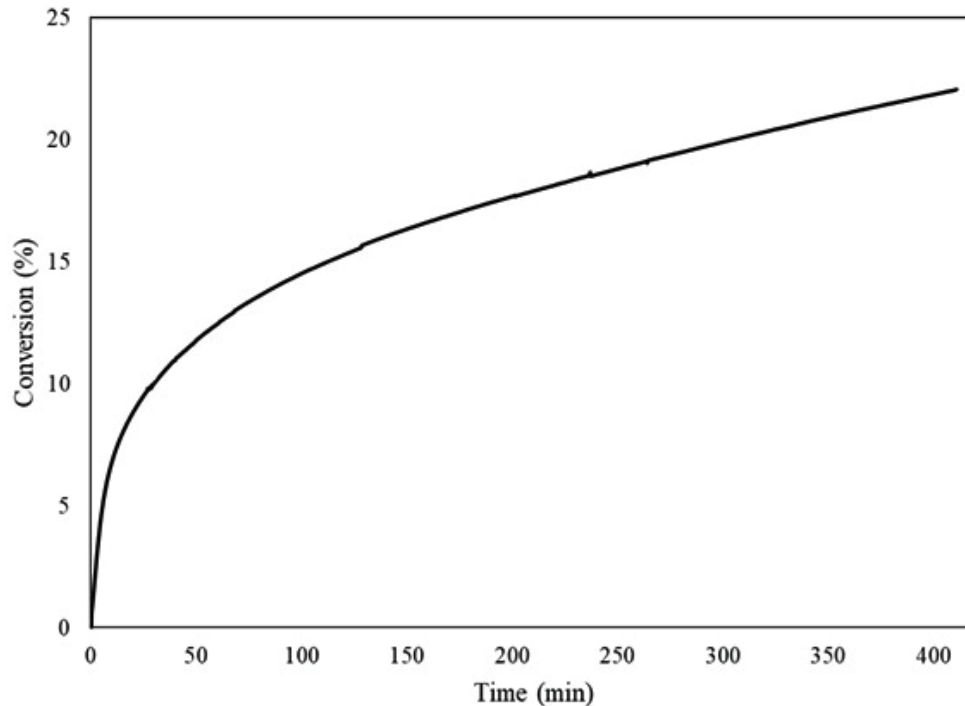


Fig. 8 – Average overall $MgSO_4$ conversion in the packed bed reactor system as the function of reaction time

vol.% with reactor length of 6 cm was about 19 min. For MgO, the breakthrough time at SO_2 concentration of 1 vol.% with reactor length of 6 cm and temperature of 500 °C was 50 min. The results of comparing breakthrough times show that MgO adsorption capacity is very much higher than that of Fe_2O_3 . The comparison of MgO rate constant with other sorbents is summarized in Table 4.

Optimization results

In this study, the performance of packed bed reactor for SO_2 removal was optimized. The optimization was carried out by NSGA II method and the Pareto-optimal solutions figure was obtained. Pareto-optimal solutions correspond to a situation in which the solutions are not dominated with respect to each other. Moving from a Pareto solution to the next one will result in a certain amount of gain in one objective and a certain amount of sacrifice in the other.

The two target objectives of the optimization were:

- to maximize the packed bed reactor breakthrough time (at $C/C_0 = 0.1$)
- to minimize the energy consumption in calcination process

Genetic algorithms are commonly used for minimizing the objective function $f(x)$. If a function is to be minimized, $-f(x)$ or $1/f(x)$ should be considered as the objective functions.

Table 4 – Summary of the comparison between MgO rate constant with other sorbents

Adsorbent	Temperature (°C)	Rate constant ($m\ s^{-1}$)
CuO	500	$4.37 \cdot 10^{-8}$
MgO		$6.01 \cdot 10^{-6}$
CaO	600	$3.12 \cdot 10^{-7}$
MgO		$1.11 \cdot 10^{-5}$

The bounds for the multi-objective optimization presented here are:

$$500\ ^\circ\text{C} \leq \text{temperature} \leq 700\ ^\circ\text{C}$$

$$1\ \text{vol.}\% \leq SO_2\ \text{concentration} \leq 2\ \text{vol.}\%$$

$$1.67 \cdot 10^{-6}\ m^3\ s^{-1}\ (100\ \text{cc}\ \text{min}^{-1}) \leq \text{zero-air flow rate} \leq 3.34 \cdot 10^{-6}\ m^3\ s^{-1}\ (200\ \text{cc}\ \text{min}^{-1})$$

The multi-objective optimization was carried out based on NSGA II. The NSGA II input parameters are presented in Table 5. The Pareto-optimal solutions after 120 generations are presented in Fig. 9 and Table 6.

The best parameters for maximizing breakthrough lifetime and minimizing the energy consumption are shown in Table 6. It can be found that improvement in breakthrough time will result in energy consumption increase. All solutions in Pareto-optimal space can be considered as the answer, but the best one has to be selected by the user based on the operating and maintenance costs, experience,

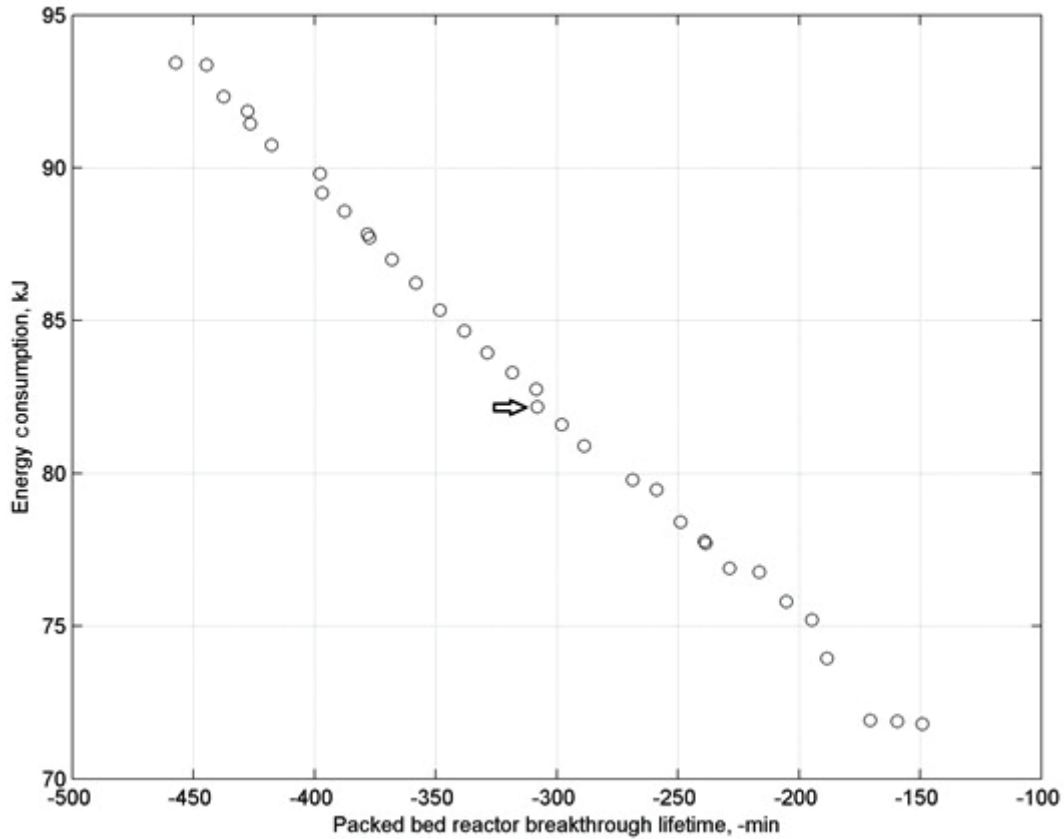


Fig. 9 – Pareto-optimal solutions

and environmental issues. According to the results presented here, the optimum SO_2 concentration, zero-air flow rate, and temperature are 1 vol.%, $1.67 \cdot 10^{-6} \text{ m}^3 \text{ s}^{-1}$ (100 cc min^{-1}), and $500\text{--}643 \text{ }^\circ\text{C}$, respectively.

The SO_2 concentration and zero-air flow rate were approximately the same in all solutions, and therefore, the appropriate temperature had to be selected. The breakthrough time of chromosomes 12 and 29 were rather similar, but the energy consumption of chromosome 12 was lower than of chromosome 29. Therefore, the situation of chromosome 12 could be considered the appropriate solution. In chromosome 12, the temperature, breakthrough time, and energy consumption were $572.55 \text{ }^\circ\text{C}$, 308.52 min , 82.76 kJ , respectively. This chromosome is shown by an arrow in Fig. 9.

Table 5 – NSGA II input parameters

Number of objectives	2
Number of bounds	3
Population size	35
Crossover method	Arithmetic crossover
Crossover probability	0.7
Mutation method	Gauss method
Mutation probability	0.05

Conclusion

In this study, SO_2 adsorption experiments were carried out in a packed bed reactor by MgO to obtain the breakthrough curves. Furthermore, the effect of important operating parameters including temperature, SO_2 concentration, and zero-air flow rate was investigated. Increasing the temperature improved the breakthrough time, but the increase in concentration and flow rate reduced the lifetime. The possibility of simultaneous production of magnesium sulfate, as a useful by product in the process of SO_2 removal by MgO, was considered. The XRD analysis verified that MgSO_4 is produced as a result of $\text{MgO} + \text{SO}_2$ reaction. The experimental results were correlated by applying RPM. The finite element method was used for solving RPM packed bed equations, and the results were compared with the obtained experimental data. There was a satisfactory agreement between the RPM values predicted by the model and the experimental breakthrough curve data.

The comparison of MgO performance with other sorbents for SO_2 removal was also carried out in this research. The reaction rate and the breakthrough time of MgO were mostly higher than other sorbents. As a result, the application of MgO in dry FGD systems could improve the efficiency of these units.

Table 6 – Pareto-optimal solutions after 120 generations

No.	Temperature (°C)	SO ₂ concentration (vol.%)	Zero-air flow rate (cc min ⁻¹)	Breakthrough time (min)	Energy consumption (kJ)
1	500	1.20	1.05	149.12	71.79
2	611.01	1.00	1.00	387.54	88.57
3	643.22	1.00	1.00	457.34	93.44
4	606.10	1.00	1.00	378.12	87.83
5	522.53	1.01	1.00	194.81	75.20
6	543.72	1.00	1.00	248.85	78.40
7	580.37	1.00	1.00	328.72	83.94
8	568.65	1.00	1.00	308.05	82.17
9	595.52	1.00	1.00	358.22	86.23
10	605.34	1.00	1.00	377.41	87.71
11	539.53	1.00	1.00	238.97	77.77
12	572.55	1.00	1.00	308.52	82.76
13	635.92	1.00	1.00	437.57	92.33
14	630.01	1.00	1.00	426.56	91.44
15	514.20	1.00	1.00	188.43	73.94
16	614.96	1.00	1.00	396.89	89.17
17	632.73	1.00	1.00	427.54	91.85
18	500	1.20	1.05	149.12	71.79
19	625.36	1.00	1.00	417.74	90.74
20	539.09	1.00	1.00	238.69	77.70
21	585.10	1.00	1.00	338.07	84.66
22	619.15	1.00	1.00	397.69	89.80
23	533.71	1.00	1.00	228.65	76.89
24	500.74	1.09	1.01	170.73	71.91
25	526.45	1.01	1.01	205.29	75.79
26	642.81	1.00	1.00	444.67	93.38
27	560.24	1.00	1.00	288.71	80.90
28	589.66	1.00	1.00	348.24	85.34
29	576.06	1.00	1.00	318.35	83.29
30	532.92	1.01	1.00	216.52	76.77
31	550.70	1.00	1.00	258.74	79.46
32	552.84	1.00	1.00	268.81	79.78
33	600.63	1.00	1.00	368.05	87.01
34	564.88	1.00	1.00	298.04	81.60
35	500.55	1.16	1.02	159.509	71.88

Finally, NSGA II was employed as a technique for multi-objective optimization to determine the best operating parameters for the removal of SO₂ by magnesium oxide in the packed bed reactor. The optimization goals were to maximize reactor breakthrough time and minimize its energy consumption. The bounds on temperature, SO₂ concentration, and zero-air flow rate were used. The Pareto-optimal solutions were obtained and the optimal operating parameters were determined. The optimum SO₂ concentration, zero-air flow rate, and temperature are 1 vol.%, 1.67·10⁻⁶ m³ s⁻¹ (100 cc min⁻¹), and 500–643 °C, respectively.

List of symbols

$a = C_A/C_{Ab}$	– Dimensionless gas concentration
$b = C_B/C_{B0}$	– Dimensionless solid concentration
C_A	– Gas concentration in the pellet, kmol m ⁻³
C_{Ab}	– Bulk gas concentration, kmol m ⁻³
C_B	– Solid reactant concentration, kmol m ⁻³
C_{B0}	– Initial solid reactant concentration, kmol m ⁻³
D_{AK}	– Knudsen diffusivity, m ² s ⁻¹
D_{AM}	– Molecular diffusivity of gas A in the pellet, m ² s ⁻¹
D_e	– Effective diffusivity of gas A in the pellet, m ² s ⁻¹
D_{e0}	– Initial effective diffusivity of gas A in pores, m ² s ⁻¹
D_p	– Diffusivity of gas A in product layer, m ² s ⁻¹
k_m	– External mass-transfer coefficient, m s ⁻¹
k_s	– Surface rate constant, m s ⁻¹
L	– Thickness of the pellet, m
L_0	– Pore length per unit volume, m ⁻²
M_B	– Molecular weight of solid reactant, kg kmol ⁻¹
M_D	– Molecular weight of solid product, kg kmol ⁻¹
r	– Pore radius, m
r_p	– Each point position in the pellet, m
\bar{r}	– Average pore radius of the pellet, m
R	– Spherical pellet radius, m
S_0	– Reaction surface area per unit volume, m ⁻¹
$Sh = \frac{k_m L}{2D_{AM}}$	– Sherwood number for external mass transfer
t	– Time, s
V_p	– Total pore volume, m ³ kg ⁻¹
x	– Axial distance from beginning of bed, m
$y = 2z/L$	– Dimensionless position in the pellet
z	– Distance from the center of the pellet, m
Z	– Ratio of molar volume of solid product to solid reactant
m_{ad}	– Mass of the adsorbent, kg

$C_{p,ad}$	– Specific heat of the adsorbent, J K ⁻¹ kg ⁻¹
T_{cal}	– Calcination temperature, K
T_0	– Environmental temperature, K
m_r	– Mass of the reactor, kg
$C_{p,r}$	– Specific heat of the reactor, J K ⁻¹ kg ⁻¹
h	– Convection coefficient, W m ⁻² K ⁻¹
A_r	– Surface area of the reactor, m ²
t_{cal}	– Calcination time, s
Q	– Required heat, J
$v_0(r)$	– Pore volume distribution function, m ² kg ⁻¹
$X(\theta)$	– Conversion of solid at each time
$\beta = 2k_s(1 - \varepsilon_0) / (v_B D_p S_0)$	– Product layer resistance
ε	– Pellet porosity
ε_0	– Initial pellet porosity
$\delta = D_e / D_{e0}$	– Variation ratio of the pore diffusion
$\theta = k_s S_0 C_{Ab} t / [C_{B0}(1 - \varepsilon_0)] = t / \tau$	– Dimensionless time
v_B	– Stoichiometric coefficient of the solid reactant
v_D	– Stoichiometric coefficient of the solid product
ρ_B	– True density of the solid reactant, kg m ⁻³
ρ_D	– True density of the solid product, kg m ⁻³
$\phi = (L/2)(k_s S_0 C_{Ab}^{n-1} / v_B D_{e0})^{1/2}$	– Thiele modulus for the pellet
ψ	– Random pore model main parameter
ζ	– Dimensionless reactor length

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