Modeling and Optimization on the Carbon Dioxide Separation from Natural Gas Using Hydrotalcite-Silica Membrane

Ahmed Daham Wiheeb*
Department of Chemical Engineering, College of Engineering, Tikrit University
chahmed@tu.edu.iq

Abstract
The process modeling and optimization of carbon dioxide (CO$_2$) separation from carbon dioxide-methane (CH$_4$) binary gas mixture through hydrotalcite (HT)-silica membrane using statistical design of experiments (DoE) is reported in this study. The effect of three important process variables, pressure difference across the membrane (100-500 kPa), temperature (30-190°C) and CO$_2$ feed concentration (10-50%) on the CO$_2$ separation performance of the membrane were investigated. The response surface methodology (RSM) coupled with central composite design (CCD) was used to build up two models to correlate the effect of process conditions to CO$_2$ permeance and CO$_2$/CH$_4$ separation selectivity. The analysis of variance (ANOVA) of the quadratic model at 95% confidence interval confirmed that the model was highly significant. The CO$_2$ feed concentration with 43% showed the best performance with a CO$_2$ permeance of 6.0x10$^{-7}$ mol.m$^{-2}$.s$^{-1}$.Pa$^{-1}$ and a CO$_2$/CH$_4$ separation selectivity of 109 at 100 kPa pressure difference across the membrane and temperature of 30°C.

Keywords: Hydrotalcite; porous membrane; carbon dioxide capture; response surface methodology.

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1. Introduction
Separation of CO$_2$ from natural gas is crucial and it has received increasing attention from researchers in energy part of the world [1, 2 & 3]. Separation of CO$_2$ increases the calorific capacity, yields better transportation conditions and prevents pipeline corrosions [4]. Microporous inorganic membranes have been widely investigated for gas separation in the past few years in comparison with polymeric membranes due to their unique thermal, chemical and mechanical stability [5, 6 & 7]. Hydrotalcite (HT) is class of anionic clays called layered double hydroxide or hydrotalcite-like compounds. HT has been intensively investigated in recent years as good adsorbents for CO$_2$ [8, 9 & 10]. The HT-silica composite membrane could have the desired features of silica and HT components. Composite membrane could exhibit high CO$_2$ adsorption capacity due to the HT, while large surface area and small pore size due to the silica component in the membrane. The high CO$_2$ adsorption capacity, large surface area and small pore size would improve the separation of CO$_2$ from natural gas [11, 12].

In the previous research studies, a novel microporous composite membrane investigated from HT material modified porous silica membrane to investigate the CO$_2$ separation from diverse gas mixtures, has been fabricated and characterized successfully [11, 12 & 13]. The modified internal pore surface of silica membrane with HT material, enhanced the adsorbed CO$_2$ amount that resulted in the increment of the CO$_2$ permeance, diffusion, and separation, respectively [14, 15]. However, very few researches reported the modeling and optimization of the process variables for gas permeation and selectivity through membrane. Here, the discussion only pertains the factorial design which was conducted by fixing all the process variables with only one variable varied at a time. The drawbacks of this technique are time consuming and difficulty to find the interaction between the process variables. Accordingly, in this paper statistical approach is applied to find the optimum operating conditions for the permeation and separation of CO$_2$ from CO$_2$/CH$_4$ binary gas mixture. Design of Experiment (DoE) is used as statistical tool to determine the optimum conditions, to evaluate the interaction between the variables and to build up an equation that can be represent the CO$_2$ permeance and CO$_2$/CH$_4$ separation selectivity and effects of surface affinity to permeability and selectivity.
2. Modeling and Optimization

2.1 Design of Experiments (DoE)

The design of experiments (DoE) was used for modeling and optimizing the permeation and separation experiments of CO2/CH4 binary gas mixture [11, 12 & 13]. Design Expert software version 6.0.6 was used in DoE. The effects of three independent variables (pressure difference across the membrane, temperature and CO2 feed concentration) on CO2 permeance and separation selectivity of CO2/CH4 were studied in the ranges of 100-500 kPa pressure difference, 30-190 °C temperature and 10-50% CO2 feed concentration. In this statistical model, 20 experimental runs were suggested by the response surface methodology (RSM) coupled with central composite design (CCD) available in the Design Expert. The main advantage of RSM is to reduce the required experimental runs required to model the permeation and separation performance of mixed gases. The Design-Expert software enables determination of the functional relationships between independent variables from minimum number of experiments. It also provides empirical model for the desired response as a function of selected variables by applying the multiple regression analysis method on the experimental data [16]. The analysis of variance (ANOVA) was implemented on the empirical model to find its statistical significance. After that, the responses were optimized by numerical optimization approach available in the Design Expert software.

The experimental design matrix of 2^3 full factorial with CCD for the permeation and separation of CO2/CH4 binary mixture using HT-silica membrane is shown in Table 1. Three factors full factorial design requires a total 20 experimental runs where it consists of 8 factorial points, 6 axial points and 6 replicates at center points. The replicate at center point (experimental runs of 15-20) was used to check the reproducibility of experimental data. The experimental runs were conducted randomly so as to minimize bias from the systematic trends in the variables. Two responses of CO2 permeance and CO2/CH4 separation selectivity were considered to study the effect of process variables. The empirical model is shown below:

\[ Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_1 x_1 x_2 + \beta_1 x_1 x_3 + \beta_2 x_2 x_3 + \beta_1 x_1^2 + \beta_2 x_2^2 + \beta_3 x_3^2 \]  

where: \( Y \) is the predicted response, \( x_1, x_2, \) and \( x_3 \) are the actual forms of pressure difference, temperature and CO2 feed concentration, respectively. The term \( \beta_0 \) is the offset, \( \beta_1, \beta_2 \) and \( \beta_3 \) are the linear terms, \( \beta_{12}, \beta_{23} \) and \( \beta_{13} \) are the interaction terms. \( \beta_1^2, \beta_2^2, \beta_3^2 \) are the quadratic terms, and \( \beta_{12}, \beta_{13}, \beta_{23} \) are the interaction terms. The analysis of variance (ANOVA) was implemented on the empirical model to find its statistical significance. After that, the responses were optimized by numerical optimization approach available in the Design Expert software.

Table 1 Experiment matrix and responses for the permeation and separation of CO2/CH4 across micro-porous HT-silica membrane.

| Run | Point Type | Pressure difference, kPa (A) | Temperature, °C (B) | CO2 feed Conc. % (C) | CO2 permeance \( \times10^{-6} \) mol/m².s.Pa (Y1) | CO2/CH4 Selectivity (Y2) |
|-----|------------|-----------------------------|---------------------|----------------------|-----------------------------------|-----------------------------|
| 1   | Part       | 180                         | 62                  | 42                   | 3.71                             | 34.25                       |
| 2   | Part       | 180                         | 158                 | 42                   | 1.86                             | 32.63                       |
| 3   | Part       | 420                         | 62                  | 10                   | 2.08                             | 23.54                       |
| 4   | Part       | 420                         | 158                 | 42                   | 0.76                             | 16.97                       |
| 5   | Part       | 180                         | 62                  | 10                   | 3.75                             | 24.44                       |
| 6   | Part       | 420                         | 158                 | 10                   | 1.67                             | 19.22                       |
| 7   | Part       | 180                         | 62                  | 42                   | 2.77                             | 19.61                       |
| 8   | Part       | 420                         | 62                  | 42                   | 1.37                             | 39.22                       |
| 9   | Axial      | 300                         | 110                 | 10                   | 2.78                             | 14.56                       |
| 10  | Axial      | 300                         | 150                 | 10                   | 3.33                             | 19.12                       |
| 11  | Axial      | 300                         | 30                  | 30                   | 3.82                             | 52.56                       |
| 12  | Axial      | 500                         | 30                  | 30                   | 3.31                             | 25.36                       |
| 13  | Axial      | 300                         | 110                 | 50                   | 0.87                             | 23.81                       |
| 14  | Axial      | 300                         | 110                 | 30                   | 4.39                             | 55.55                       |
| 15  | Center     | 300                         | 110                 | 30                   | 1.65                             | 29.76                       |
| 16  | Center     | 300                         | 30                  | 30                   | 2.19                             | 31.96                       |
| 17  | Center     | 300                         | 110                 | 30                   | 2.16                             | 30.54                       |
| 18  | Center     | 300                         | 110                 | 50                   | 1.82                             | 22.18                       |
| 19  | Center     | 300                         | 110                 | 30                   | 1.78                             | 27.72                       |
| 20  | Center     | 300                         | 110                 | 30                   | 2.15                             | 26.56                       |

2.2 Gas permeation and separation selectivity

The measured permeance of gas species \( i \) in the binary gas mixture, \( K_i \) was obtained by taking the ratio of mole flux of the gas species \( i \), \( N_i \), in the permeate to the log-mean pressure difference, \( \Delta P_{\text{log}} \), across the membrane,

\[ K_i = \frac{N_i}{\Delta P_{\text{log}}} \]  

where,

\[ \Delta P_{\text{log}} = \frac{(p_{F,i} - p_{P,i}) - (p_{R,i} - p_{P,i})}{\ln((p_{F,i} - p_{P,i})/(p_{R,i} - p_{P,i}))} \]  

\( p_{F,i} \) and \( p_{R,i} \) are the partial pressures for gas species \( i \) in feed, permeate and retentate, respectively. The separation selectivity of binary mixture, \( a_{ij} \), between 2 gases can be computed based on the ratio of the permeance of gas species-\( i \) to that of species-\( j \) in the binary mixture containing of species \( i,j \).
$$a_i = K_i/K_j \quad (4)$$

3. Results & Discussion

3.1 Response surface modeling of CO₂ permeance

Table 2 presents the statistical results of the analysis using ANOVA for the CO₂ permeance in the CO₂/CH₄ binary mixture. The ANOVA results show that the CO₂ permeance is best described with a polynomial model. This quadratic model is proposed by RSM software in terms of actual factors, as shown in Equation 5. The positive sign (+) in front of model terms designates synergistic effect, increase the CO₂ permeance, whereas the negative sign (-) designates antagonistic effect, decrease the CO₂ permeance.

$$K_{CO_2} = +8.6698 - 0.0206A - 0.0361B + 3.8567C + 1.8559 \times 10^{-5}A^2 + 8.5526 \times 10^{-5}B^2 - 5.0659C^2 + 3.9996 \times 10^{-5}AB - 5.9220 \times 10^{-5}AC - 0.0236BC \quad (5)$$

ANOVA is applied for estimation of the main variable effects and their potential interaction on the CO₂ permeance. The most important outputs from ANOVA results are the Fisher value (F-value) and associated probability value (Prob>F). The (Prob>F) value indicates the probability equals the proportion of the area under the curve of the F-distribution that lies beyond the observed F-value. In other words, when the (Prob>F) value is small, the particular term is considered to significantly affect the CO₂ permeance. The model terms with (Prob>F) less than 0.05 indicate that the terms are significant for the model of CO₂ permeance. The model F-value is 25.72 and Prob>F value is less than 0.05 implying that the developed model is significant at 95% confident level. In present study, terms A, B, C, A², B² and AB are significant for CO₂ permeance at 95% confident level. Although the terms C², AC and BC are not significant to the model because of their values of probability (Prob>F) are greater than 0.05, they are involved in Equation 5 to get a hierarchy model [16]. The lack of fit is the ratio between the residuals and pure error. Lack of fit F-value of 3.11 with (Prob>F) of 0.1194 denotes that it is not significant comparative to the pure error due to noise, consequently the suggested model for CO₂ permeance in Equation 5 is valid for the present study.

Based on the ANOVA result in Table 2, the F-values show the pressure difference across the membrane gives the greater impact on CO₂ permeance followed by temperature and CO₂ feed concentration. The interaction between pressure difference and temperature on CO₂ permeance at central level of CO₂ feed concentration (30%) is shown in the three-dimensional response surface plot in Figure 1. As it can be observed from the figure that the CO₂ permeance decreases with the increase of pressure difference from 100-500 kPa for separation temperature ranging from 30 to 190 °C. This result is consistent with those reported in literature for the membranes that considered surface diffusion mechanism for CO₂ separation [17]. The CO₂ permeance declined because the coverage-gradient driving force increase is less than the CO₂ partial pressure difference. As the pressure difference across the membrane is increased, the adsorbed concentration gradient of CO₂ increases, and therefore, CO₂ flux increases. The increase in CO₂ flux with increase in the adsorbed concentration gradient is lower than the increase in CO₂ partial pressure difference [18, 19]. Then, the CO₂ permeance (CO₂ flux/ The log-mean partial pressure difference) declines as the pressure difference is increased from 100 to 500 kPa. At low temperature (30 °C) the developed model predicts that the CO₂ permeance declined from 6.22×10⁻⁷ to 2.20×10⁻⁷ mol.m⁻².s⁻¹.Pa⁻¹ as the pressure difference is increased from 100-500 kPa, while at high temperature from 2.96×10⁻⁷ to 1.50×10⁻⁷ mol.m⁻².s⁻¹.Pa⁻¹. At high temperature the CO₂ adsorption coverage on HT is low. Thus, the CO₂ permeance decreases with increase in the pressure difference at high temperature less than at low temperature. The response surface plot of CO₂ permeance at 110 °C with a different CO₂ feed concentration and pressure difference is shown in Figure 2. Generally the trend of CO₂ permeance is rather similar to that of Figure 3. The increase in the separation temperature leads to reduction of the surface coverage and at the same time increases the CO₂ micro pore diffusivity. Hence, the CO₂ permeance decreases with increase in separation temperature, which means the surface coverage promotes the micro pore diffusivity [19]. The CO₂ permeance decreases with increase in CO₂ feed concentration from 10 to 50% because the coverage-gradient driving force increase is less than the CO₂ partial pressure difference [20].
Table 2 ANOVA results for CO\(_2\) permeance in the CO\(_2\)/CH\(_4\) binary mixture.

| Source | Sum of DF \*(Sum| Mean | F-value | P-value |
|--------|----------------|------|---------|---------|
| Total  | 18.33 | 9 | 2.09 | 25.72 | <0.0001 |
| A      | 0.06  | 1 | 0.06 | 111.75 | <0.0001 |
| B      | 4.73  | 1 | 4.73 | 59.42  | <0.0001 |
| C      | 2.45  | 1 | 2.45 | 30.75  | <0.0001 |
| A\(^2\) | 0.09  | 1 | 0.09 | 12.47  | 0.0054 |
| B\(^2\) | 0.24  | 1 | 0.24 | 6.78   | 0.0063 |
| C\(^2\) | 0.074 | 1 | 0.074| 0.90   | 0.3700 |
| AB     | 0.41  | 1 | 0.41 | 5.14   | 0.0467 |
| AC     | 0.226 | 1 | 0.226| 0.70   | 0.4380 |
| BC     | 0.14  | 1 | 0.14 | 1.80   | 0.2097 |
| Residual | 0.00  | 10 | 0.00 | -     | -      |
| Lack of Fit | 0.60  | 5 | 0.12 | 3.11  | 0.1194 |
| Pure Error | 0.19  | 5 | 0.039| -    | -      |
| Cor Total | 19.23 | 19 | -    | -    | -      |

R\(^2\) = 0.9596

\*DF = Degree of Freedom, \*\* = Standard Deviation

Figure 1: Effect of pressure difference across the HT-silica membrane and temperature on CO\(_2\) permeance at 30\% CO\(_2\) feed concentration

Figure 2: Effect of pressure difference across the HT-silica membrane and CO\(_2\) feed concentration on CO\(_2\) permeance at temperature of 110ºC

Figure 3: Effect of temperature and CO\(_2\) feed concentration on CO\(_2\) permeance at pressure difference across the HT-silica membrane of 300 kPa

3.2 Response surface modeling of CO\(_2\)/CH\(_4\) separation selectivity

The analysis of variance ANOVA for CO\(_2\)/CH\(_4\) separation selectivity is presented in Table 3. The ANOVA results show that the CO\(_2\)/CH\(_4\) separation selectivity is best presented with a quadratic model. This model is suggested by RSM software in terms of actual factors, as shown in Equation 6.
As shown in ANOVA Table 3, F-value of the developed model gives a value of 32.11 with Prob>F value of <0.05 implying that the model of CO\textsubscript{2}/CH\textsubscript{4} separation selectivity is significant at 95% confidence level. In this study, A, B, C, A\textsuperscript{2}, B\textsuperscript{2}, C\textsuperscript{2} and AB are significant terms for CO\textsubscript{2}/CH\textsubscript{4} separation selectivity. Whereas AC and BC are not significant to the CO\textsubscript{2}/CH\textsubscript{4} separation selectivity because of their values of probability (Prob>F) are higher than 0.05. However, they are included in Equation 6 to get a hierarchy model [16]. Lack of fit F-value of 3.77 with (Prob>F) of 0.0858 implies that it is not significant relative to the pure error due to noise. As a result, the proposed model for CO\textsubscript{2}/CH\textsubscript{4} separation selectivity in Equation 6 is valid for the present study.

According to the F-values of the ANOVA result in Table 3, the temperature shows the highest effect on CO\textsubscript{2}/CH\textsubscript{4} separation selectivity followed by the pressure difference across the HT-silica membrane and CO\textsubscript{2} feed concentration. The quadratic effect of pressure difference (A\textsuperscript{2}), temperature (B\textsuperscript{2}), CO\textsubscript{2} feed concentration (C\textsuperscript{2}) and the interaction pressure difference and temperature (AB) are relatively significant, with the F-value of 15.56, 7.57, 8.5 and 9.82, respectively. The dimensional response surface plots for the CO\textsubscript{2}/CH\textsubscript{4} separation selectivity at center level with its interaction between pressure difference, temperature and CO\textsubscript{2} feed concentration are shown in Figure 4, Figure 5 and Figure 6. It can be seen from Figure 6 that the CO\textsubscript{2}/CH\textsubscript{4} separation selectivity decreased as the pressure difference was increased from 100-500 kPa. The CO\textsubscript{2} flow mechanism is surface diffusion and micropore diffusion due to the high CO\textsubscript{2} adsorption capacity (47.48 mg CO\textsubscript{2}/g sorbent) and small pore diameter (8Å) of the HT-silica membrane [12], while CH\textsubscript{4} flow mechanism is micropore diffusion. The weakly adsorbed molecule CH\textsubscript{4} was hindered from penetrating through the micro-porous HT-silica membrane due to mouth narrowing by adsorbed CO\textsubscript{2} molecules [19-21]. The CO\textsubscript{2} permeance decreases proportionally more than CH\textsubscript{4} permeance, and thus the CO\textsubscript{2}/CH\textsubscript{4} selectivity is decreased. The movement of the gases molecules through membrane pores was affected by the intermolecular collision, the interaction between gas molecules and membrane pore wall, and the interplay between the movements of different gas molecules and its relation with the kinetic diameter of the gas molecule [17, 19]. The effect of CO\textsubscript{2} feed concentration on CO\textsubscript{2}/CH\textsubscript{4} separation selectivity at different pressure difference and temperature are shown in Fig 5 and Figure 6. It can be observed that the CO\textsubscript{2}/CH\textsubscript{4} separation selectivity increases with increase in CO\textsubscript{2} feed concentration due to the increase of CO\textsubscript{2} loading and more CO\textsubscript{2} molecules available in the gas mixture leads to increase the blocking effect and decrease the CH\textsubscript{4} permeance. However, at CO\textsubscript{2} feed concentration near saturation, increasing the CO\textsubscript{2} feed concentration caused a small increase in CO\textsubscript{2} loading along the membrane wall, thus the CH\textsubscript{4} permeance remained constant while the CO\textsubscript{2} permeance continued to decrease with increase in CO\textsubscript{2} feed concentration [17, 20].

\[ \alpha_{\text{CO}_2/\text{CH}_4} = +86.0929 - 0.3023 A - 0.6162 B + 274.4279 C + 2.9250 \times 10^{-4} A^2 + 1.2711 \times 10^{-3} B^2 - 215.5016 C^2 + 7.7738 \times 10^{-4} AB - 0.1476 AC - 0.4866 BC \]  

**Table 3** ANOVA results for separation selectivity in the CO\textsubscript{2}/CH\textsubscript{4} binary mixture.

| Source | Sum of DF\* Squares | Mean DF\* Value | Prob > F |
|--------|---------------------|-----------------|----------|
| Model  | 4551.71             | 9               | 32.11    | <0.0001 | Significant |
| A      | 1419.44             | 1               | 1419.44  | <0.0001 |            |
| B      | 1920.70             | 1               | 1920.70  | <0.0001 |            |
| C      | 432.48              | 1               | 432.48   | 0.0034  |            |
| A\textsuperscript{2} | 246.59             | 1               | 246.59   | 0.0027  |            |
| B\textsuperscript{2} | 115.21              | 1               | 115.21   | 0.0204  |            |
| C\textsuperscript{2} | 133.85              | 1               | 133.85   | 0.0134  |            |
| AB     | 154.70              | 1               | 154.70   | 0.0160  |            |
| AC     | 58.86               | 1               | 58.86    | 0.1877  |            |
| BC     | 60.61               | 1               | 60.61    | 0.0782  |            |
| Residual | 157.52            | 10              | 15.75    |         |            |
| Lack of Fit | 124.49            | 5               | 24.90    | 0.0859  | Not significant |
| Pure Error | 33.03             | 5               | 6.61     |         |            |
| Cor Total | 4709.23           | 19              |         |         |            |
| **DF** | 19                 |                  |          |         |            |
| **R\textsuperscript{2}** | 0.9966          |                  |          |         |            |

DF* = Degree of Freedom, *R** = Standard Deviation

**Figure 4:** Effect of pressure difference across the HT-silica membrane and temperature on CO\textsubscript{2}/CH\textsubscript{4} separation selectivity at 30% CO\textsubscript{2} feed concentration
3.3 Process optimization using response surface methodology (RSM)

The eventual objective of RSM is to find out the optimum conditions as compromise between higher CO₂ permeance and CO₂/CH₄ separation selectivity. The optimum condition preset at high and low level ranges of the three independent variables: pressure difference, temperature and CO₂ feed concentration for separation studies of CO₂/CH₄ binary gas mixture over HT-silica membrane is obtained using numerical optimization feature of Design Expert 6.0.6 Software. The optimization module in Design Expert searches for a combination of variable levels that simultaneously satisfy the requirements placed for all the responses and variables. Table 4 summarizes the optimization criteria used to seek the optimum values for the two responses. Table 5 presents the four optimum solutions generated by Design Expert software. By default, the solutions are sorted from best to worst depending on total desirability. In the present work, the desirability function approach is used in the response surface methodology to optimize the operating conditions. The experimental conditions that provide the highest desirability response value are determined using this method. The optimum conditions in Solution 1 with the highest total desirability of 1.000 are chosen for further process studies. The optimum conditions for CO₂ permeance and CO₂/CH₄ separation selectivity are found to be 100.94 kPa pressure difference, 30.09 °C separation temperature and 43.05% CO₂ feed concentration. The developed model predicted that an optimum CO₂ permeance of 6.0×10⁻⁷ mol.m⁻².s⁻¹.Pa⁻¹ and CO₂/CH₄ separation selectivity of 109 can be obtained.

In order to check the validity of the DoE model prediction, the predicted optimum for all the responses are then verified by performing five repeated experiments at the optimum conditions (solution 1). Table 6 summarizes the separation results for the CO₂/CH₄ binary gas mixture of the repeated experimental runs. The experimental values of the CO₂ permeance and CO₂/CH₄ separation selectivity at the optimum conditions are compared with the predicted values by the DoE software. The mean% error is 3.27 for CO₂ permeance and 3.80 for CO₂/CH₄ separation selectivity. The mean% error for both responses is less than 5% means that the experimental value is close to the predicted value. It can be concluded that the developed model by DoE software with RSM in this study shows good predictability and sufficient reliability for the modeling and predicting the HT-silica membrane performance for the permeation and separation of CO₂/CH₄ binary gas mixture.

| Criteria                  | Goal                  | Lower limit | Upper limit |
|---------------------------|-----------------------|-------------|-------------|
| Pressure Difference, kPa  | In the range          | 10          | 500         |
| Temperature, °C           | In the range          | 30          | 190         |
| CO₂ Feed Concentration, % | In the range          | 10          | 50          |
| Permeance x10⁻⁷, mol.m⁻².s⁻¹.Pa⁻¹ | Maximum             | 0.76        | 5.21        |
| CO₂/CH₄ Separation Selectivity | Maximum             | 5           | 106         |

Table 4  Constraint used for optimization of CO₂ permeance and separation selectivity of the CO₂/CH₄ binary gas mixture.
Table 5 Optimum condition of CO₂ permeance and separation selectivity of the CO₂/CH₄ binary gas mixture.

| Run | Pressure difference (atm) | Temperature (°C) | CO₂ Feed % | CO₂ Permeance (×10⁻⁹ mol·m⁻²·s⁻¹·Pa⁻¹) | CO₂ Selectivity | CO₂/CH₄ Separation Selectivity (%) |
|-----|--------------------------|------------------|-------------|----------------------------------------|-----------------|-------------------------------------|
| 1   | 100.94                   | 30.99            | 6.0         | 1.142                                  | 108.9           | 4.97                                |
| 2   | 101.04                   | 30.94            | 6.0         | 1.116                                  | 108.9           | 5.30                                |
| 3   | 100.00                   | 30.00            | 6.0         | 1.104                                  | 108.9           | 5.56                                |

Table 6 Experiments at optimum conditions simulated by DoE for the CO₂ permeance and separation selectivity of CO₂/CH₄ binary gas mixture.

| Run | CO₂ Permeance (×10⁻⁹ mol·m⁻²·s⁻¹·Pa⁻¹) | % Error | CO₂ Selectivity | % Error | CO₂/CH₄ Separation Selectivity (%) | % Error |
|-----|----------------------------------------|--------|----------------|---------|------------------------------------|--------|
| 1   | 6.13                                  | 1.37   | 108.9          | 4.97    |                                    | 3.80   |
| 2   | 6.21                                  | 1.39   | 108.9          | 5.30    |                                    | 3.80   |
| 3   | 6.29                                  | 1.37   | 108.9          | 5.56    |                                    | 3.80   |
| 4   | 5.94                                  | 1.37   | 108.9          | 5.56    |                                    | 3.80   |
| 5   | 6.17                                  | 1.37   | 108.9          | 4.40    |                                    | 3.80   |

4. Conclusions

Process modeling and optimization of mixed gas permeation and separation of CO₂/CH₄ through HT-silica membrane were performed using RSM coupled with CCD available in statistical method of DoE. The analysis of the RSM showed that the process variables (pressure difference, operating temperature and CO₂ feed concentration) had significant effects on CO₂ permeance and CO₂/CH₄ separation selectivity. From the study, it was evident that optimal conditions that maximize CO₂ permeance and favor higher CO₂/CH₄ separation selectivity were unfavorable at higher temperature and pressure difference. The developed models by DoE showed good predictability and sufficient reliability for the modeling and predicting the CO₂ permeance and separation selectivity of CO₂/CH₄ mixed gas through HT-silica membrane with the values of correction coefficient (R²) higher than 0.95.

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