Gypsum Scale Formation and Inhibition Kinetics with Implications in Membrane System

Zhaoyi Dai ^{1, 2, 3,*}, Yue Zhao ^{3, 4}, Samridhdi Paudyal ³, Xin Wang ³, Chong Dai ³, Saebom Ko ³, Wei Li³, Amy T. Kan³, Mason B. Tomson³

¹ State Key Laboratory of Biogeology and Environmental Geology, School of Earth Sciences, China University of Geosciences, Wuhan 430074, China

² Hubei Key Laboratory of Critical Zone Evolution, School of Earth Sciences, China University of Geosciences, Wuhan 430074, China

³ Department of Civil and Environmental Engineering, Rice University, 6100 Main Street,

Houston, Texas 77005, United States

⁴ Research Institute of Petroleum Processing, SINOPEC, Beijing, China

* Corresponding author. Tel.: +86-13675620524, E-mail: dzy616@gmail.com

Abstract

 Water desalination using membrane technology is one of the main technologies to resolve water pollution and scarcity issues. In the membrane treatment process, mineral scale deposition and fouling is a severe challenge that can lead to filtration efficiency decrease, permeate quality compromise, and even membrane damage. Multiple methods have been developed to resolve this problem, such as scale inhibitor addition, product recovery ratio adjustment, periodic membrane surface flushing. The performance of these methods largely depends on the ability to accurately predict the kinetics of mineral scale deposition and fouling with or without inhibitors. Gypsum is one of the most common and troublesome inorganic mineral scales in membrane systems, however, no mechanistic model is available to accurately predict the induction time of gypsum crystallization and inhibition. In this study, a new gypsum crystallization and inhibition model based on the classical nucleation theory and a Langmuir type adsorption isotherm has been developed. Through this model, it is believed that gypsum nucleation may gradually transit from homogeneous to heterogeneous nucleation when the gypsum saturation index (*SI*) decreases. Such transition is represented by a gradual decrease of surface tension at smaller *SI*. This model assumes that the adsorption of inhibitors onto the gypsum nucleus can increase the nucleus superficial surface tension and prolong the induction time. Using the new model, this study accurately predicted the gypsum crystallization induction times with or without nine commonly used scale 19 inhibitors over wide ranges of temperature $(25 - 90 \degree C)$, *SI* (0.04 – 0.96), and background NaCl 20 concentration $(0 - 6 \text{ mol/L})$. The fitted affinity constants between scale inhibitors and gypsum show a good correlation with those between the same inhibitors and barite, indicating a similar inhibition mechanism via adsorption. Furthermore, by incorporating this model with the two-phase mineral deposition model our group developed previously, this study accurately predicts the gypsum deposition time on the membrane material surfaces reported in the literature. We believe that the model developed in this study can not only accurately predict the gypsum crystallization induction time with or without scale inhibitors, elucidate the gypsum crystallization and inhibition mechanisms, but also optimize the mineral scale control in the membrane filtration system.

Keywords

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1. Introduction

 Water scarcity is becoming more severe due to the continuously growing population, developing economies, and increasing water pollutions (Dai and Li, 2013; Huang et al., 2020; Yu et al., 2019). Water desalination has been playing a more important role especially when cost continues to drop (Ghaffour et al., 2013; Huang et al., 2020). Ghaffour et al. (2013) reported that the membrane process accounts for 63.7% of the total global desalted water capacity. In membrane processes, the elevation of ion concentrations, the changes of temperature and pressure, and the occurrence of concentration polarization can lead to the mineral scale formation and fouling in the membrane system (Al-Roomi and Hussain, 2016; Lioliou et al., 2006; Warsinger et al., 2015; Yu et al., 2019). Such inorganic scale formation has become a major limiting factor of membrane system applications (Lin et al., 2005; Rabie et al., 2001; Seidel and Elimelech, 2002; Speth et al., 1998; Van der Bruggen et al., 2001). Specifically, it can decrease the permeation rate, increase the osmotic pressure, increase contaminant breakthrough and compromise permeate water quality, and cause membrane damage (Huang et al., 2020; Lin et al., 2005; Shirazi et al., 2010; Warsinger et al., 2015).

 Gypsum (CaSO4·2H2O, calcium sulfate dihydrate) is one of the most common inorganic mineral scales in membrane systems (Warsinger et al., 2015; Yu et al., 2019). Since gypsum is not acid soluble and is a tenacious and adherent scale on the membrane surfaces, it has become one of the most troublesome scale minerals (Gryta, 2009). To effectively control the gypsum scale problem, different strategies have been proposed to ensure the induction time is longer than the crystallization or deposition time in the membrane operation system using different techniques. 58 For example, the cations (i.e., Ca^{2+}) and/or anions (i.e., $SO_4{}^{2-}$) can be selectively removed to decrease the saturation level of gypsum (and thus increase the induction time) by ion-exchange (Kelle Zeiher et al., 2003; Mukhopadhyay, 2003) or capacitive deionization process using composite electrodes (Zuo et al., 2018). Antony et al. (2011) suggested to limit the product recovery so that the saturation level of gypsum in the concentrate stream is controlled below a certain saturation level, which limited the economic efficiency of membrane system at the same time (Antony et al., 2011). It was also proposed to periodically flush membrane surface or reverse flow directions to shorten the time for crystallization and deposition less than the induction time (Lee et al., 2018; Nghiem and Cath, 2011; Pomerantz et al., 2006; Uchymiak et al., 2009; Warsinger et al., 2015). At the same time, various scale inhibitors (or antiscalants) have been widely used to prolong the induction time (Al-Roomi and Hussain, 2016; Antony et al., 2011; Lin and Singer, 2005; Lioliou et al., 2006; Rosenberg et al., 2012; Warsinger et al., 2015; Yu et al., 2019). Among all these methods above, the accurate prediction of the gypsum crystallization induction time is critical to achieve optimal efficiency.

 The mineral induction time (*tind*) is defined as the time elapsed between the establishment of supersaturation and the detection of mineral formation (Mullin, 2001; Nielsen, 1964). This definition shows that the induction time has a significant practical meaning and it has been widely used to indicate when the damage starts to occur in different processes (He et al., 1996; Knezic et al., 2004; Mullin, 2001; Nielsen, 1964; Van der Leeden et al., 1992). In previous research, different

 experimental methods were utilized to measure the gypsum induction time, including measuring solution turbidity with a turbidity meter or a laser apparatus (He et al., 1994a, 1994b; Lancia et al., 1999; Paudyal et al., 2020), measuring partial pressure change in a dynamic scale loop (Al-Roomi and Hussain, 2016), or measuring the frequency change on a QCM (quartz crystal microbalance) sensor (Alimi et al., 2003). Yan et al. (2015) and Zhang et al. (2017) have confirmed that the induction time achieved by monitoring solution turbidity and partial pressure of tubing are comparable with each other (Yan et al., 2015; Zhang et al., 2017). Using these experimental apparatus, the induction time of gypsum crystallization without inhibitors (Alimi et al., 2003; He et al., 1994a) and with various inhibitors (Al-Roomi and Hussain, 2016; Lioliou et al., 2006; Paudyal et al., 2020; Prisciandaro et al., 2006; Rosenberg et al., 2012; Yu et al., 2019) have been determined. Unfortunately, there is not a quantitative model available that can accurately predict gypsum induction time with or without scale inhibitors at different operation conditions. He and his colleagues developed semi-empirical models for the induction time prediction of calcite, barite, and celestite with or without scale inhibitors (He et al., 1999, 1996, 1995a, 1995b, 1994b). However, the induction times of gypsum does not show the similar trend with other minerals and cannot be described by the semi-empirical models, probably due to the relatively larger solubility and variable surface tension of gypsum at different saturation levels (He et al., 1994a). It was also suggested that the fundamentals of inhibition mechanisms are not well understood with little guidance from literature (Al-Roomi and Hussain, 2016). Therefore, in order to optimize the gypsum scale control in the membrane system with the methods introduced above, a quantitative model with solid theoretical basis is needed to accurately predict the gypsum crystallization induction time with or without inhibitors.

 In this study, a new model based on classical nucleation theory (CNT) with saturation index (*SI*) dependent surface tension is developed to predict the induction time of gypsum crystallization. This mechanistic model shows that the surface tension decreases when the gypsum crystallization transits from homogeneous nucleation at relatively larger *SI* values to heterogeneous nucleation at smaller *SI* values. A Langmuir-type adsorption model was adopted to quantify the impacts of scale inhibitors. This new model assumes that the inhibition mechanism is dominated by the inhibitor adsorption to the nucleus surface and the alternation of the effective surface tension, and can accurately predict the gypsum crystallization induction time with or without nine commonly used 107 scale inhibitors over wide ranges of temperature $(25 - 90 \degree C)$, *SI* (0.04 – 0.96), and background NaCl concentration (0 – 6 mol/L). Such good agreement between the model predictions and the experimental data supports the validity of this new model. This study also accurately predicts the gypsum deposition time on membrane material surfaces that is reported in the literature by incorporating the deposition model our group developed previously. Using this new model with solid theoretical basis, the gypsum scale management strategies can be optimized, such as recovery rate, flow reversal frequencies, inhibitor dosages, to mention a few. Such optimizations can significantly increase membrane operation efficiencies, drop the water desalination cost, and promote the wider applications of membrane technologies, as well as other technologies wherein gypsum formation is problematic.

2. Materials and Methods

2.1.Induction Time Testing

 Reagent grade CaCl2·2H2O, Na2SO4, and NaCl salts were used to prepare solutions. The cation (CAT) and anion solution (AN) solutions are prepared separately to include Ca^{2+} and SO_4^{2-} ions at designed concentrations (Table S1), respectively. The saturation index of gypsum, defined as the base ten (Briggsian) logarithms of the ion activity product over the solubility product, after mixing the CAT and AN solutions with equal volume were calculated using *ScaleSoftPitzer* (*SSP*) model (Dai et al., 2017a, 2014; Kan et al., 2015; Kan and Tomson, 2012), which is widely used as one of most accurate thermodynamic models (Mavredaki et al., 2011). In [Figure 1](#page-5-0) is plotted the multi-channel laser testing apparatus for induction time testing designed by Paudyal et al. (2020) (Paudyal et al., 2020). This apparatus is modified from the single channel laser testing apparatus used in previous research (C. Dai et al., 2019; Dai et al., 2017b; He et al., 1995a; Yan et al., 2015). The customized heating block (Chemglass, CG-1991-03) has seven reaction vial (with a volume of 30 mL) holders and seven tests can be conducted simultaneously. The laser light path was drilled in the heating block with a cross-section view shown in [Figure 1.](#page-5-0) CAT and AN solutions were added at the same time into the reaction vial using two pipettes and the recording of induction time was started. The heating and stirring is conducted by a digital hot plate stirrer (MR Hei-Tec).The induction time is picked when the laser signal starts to drop by about 1%, which indicates the formation of detectable minerals (Z. Dai et al., 2019). Various commonly used scale inhibitors were added with different dosages in the AN solution to evaluate the inhibitor impacts. The tested inhibitors are listed in [Table 1](#page-6-0) and their structures are plotted in [Figure 2.](#page-7-0) The relative errors of the induction time measurement are around 5%.

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- Figure 1. Top (left) and side (right) view of the schematic diagrams of the multi-channel laser
- testing apparatus.
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145 Table 1. Information of the inhibitors used in this study.

^{*} Molecular weight of monomer is used since the molecular weight of the polymer is not available.

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- Figure 2. Structures of scale inhibitors.

152 Table 2. The experimental conditions of the precipitation and inhibition kinetics for gypsum.

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155 **2.2.Gypsum Crystallization and Inhibition Model Development**

156 The chemical reaction of gypsum crystallization and dissolution can be described as:

157 $CaSO_4 \cdot 2H_2O \rightleftharpoons Ca^{2+} + SO_4^{2-} + 2H_2O$ (1)

 The saturation index of gypsum, *SI*, which is defined as the base ten logarithm of the ion 159 activity product (*IAP*) over the solubility product: $SI_{avpsum} = log_{10}(IAP/K_{sp.avpsum}) =$ $\log_{10}(a_{Ca^{2+}}a_{SO_4^2} - a_{H_2O}^2/K_{sp})$. In this equation, *a* represents the activity of each species and equals to the product of the concentration (mol/kg H2O, *m*) and activity coefficient (*γ*); *Ksp* is the gypsum solubility product. *SI* equals to 0 when the mineral is in equilibrium with the aqueous solution, positive when the mineral is supersaturated and has the potential to precipitate, and negative when the mineral is undersaturated. The *SI* values of each mineral and activity coefficient of each species at different conditions are calculated by the *SSP* model.

 The supersaturation of minerals will lead to mineral crystallization with the decrease of Gibbs free energy and the increase of interfacial energy (Mullin, 2001). [Söhnel and Mullin \(1979,](file:///C:/Users/zd6/Dropbox/Research/2017%20SPE/Induction%20time%20model_Joey_2.docx%23_ENREF_63) [1988\)](file:///C:/Users/zd6/Dropbox/Research/2017%20SPE/Induction%20time%20model_Joey_2.docx%23_ENREF_63) assumed that the mineral crystallization started with nucleation followed by polynuclear growth in the crystal growth stage, and the induction times were derived as follows (Nielsen, 1964; Söhnel and Mullin, 1988):

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$$
t_{ind} = \left(\frac{3}{2\pi}\right)^{1/4} \left(\frac{V_m^{5/3}}{A_v^{8/3}D^4c_{eq}}\right)^{1/4} \left(\frac{10^{SI/2}}{(10^{SI/2}-1)^2}\right)^{1/4} \exp\left(\frac{\beta V_m^2 \sigma^3 A_v}{4(RT)^3 (2.303 \cdot SI)^2} + \frac{\beta' V_m^{4/3} \sigma^2 A_v^{2/3}}{4(RT)^2 (2.303 \cdot SI)}\right)
$$
\n173 , (2)

174 where *R* is the ideal gas constant (8.31 J K⁻¹ mol⁻¹); *T* is temperature in Kelvin (K); V_m is the molar 175 volume of mineral; σ is the superficial interfacial energy between mineral and solution (J m⁻²); A_v 176 is the Avogadro constant $(6.02 \times 10^{23} \text{ mol}^{-1})$; *D* is the effective diffusion coefficient (m² s⁻¹); *c_{eq}* is 177 the equilibrium concentration of the lattice ions (mol m^{-3}) that can be calculated using *ScaleSoftPitzer*; β and β ' are the shape factors (i.e., for spheres, $\beta = 16\pi/3$, $\beta' = \pi$); the first term in the exponential part is the contribution of nucleation and the second term is due to crystal growth. He et al. (1994a) found that the logarithm of the gypsum crystallization induction time does not 181 follow a simple linear relationship with ST^2 , as suggested by CNT (He et al., 1994a). Such deviation from the CNT was proposed to be because gypsum has two different effective surface tensions, one for heterogeneous and one for homogenous nucleation (Mullin and Ang, 1976; Söhnel and Mullin, 1988, 1978). However, this assumption did not specify the transition *SI* value between these two types of nucleation and could lead to an abrupt change of induction time around the transition *SI*, which was inconsistent with the experimental observations.

 In this study, it is assumed that the gypsum nucleation is dominated by the homogeneous nucleation when the *SI* is large (Huang et al., 2020; Lancia et al., 1999; Mullin, 2001). When the gypsum *SI* becomes smaller (i.e., less supersaturated), the nucleation process is gradually more dominated by heterogenous nucleation that may happen on the surface of impurities. Such transition can be represented by a gradual decrease of surface tension using the following sigmoid type of function:

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$$
\sigma = \sigma_0 \times (1 + \lambda_T T) \times \left(1 + \frac{a}{1 + e^{-b(SI - c)}}\right).
$$
 (3)

194 where σ_0 , *a*, *b*, and *c* are constants to be fitted; λ_T is the temperature dependence coefficient of surface tension. It deserves notice that this sigmoid function is selected only to ensure a smooth transition between the homogenous- and heterogenous-dominated nucleation. The calculated value represents the superficial surface tension changing with the supersaturation of gypsum.

 It is proposed that the scale inhibitors will adsorb onto the surface of gypsum nucleus and thereby change the surface tension of the nucleus. This mechanism was assumed to be the dominant mechanism in this model. The adsorption can be described by a Langmuir type adsorption isotherm as follows:

$$
q_e = \frac{q_{\rm m} K_L [Inh]}{1 + K_L [Inh]}
$$
\n
$$
(4)
$$

203 where q_e is the effective inhibitor concentration on the nucleus surface (mol/m² nucleus surface); 204 q_m is the maximum inhibitor concentration on the surface (mol/m² nucleus surface). Divide both 205 sides by q_m :

$$
\theta = \frac{K_L[Inh]}{1 + K_L[Inh]}
$$
\n⁽⁵⁾

207 where $\theta = q_e/q_m$ and is the effective fraction coverage of the surface-active sites. The adsorption of inhibitors onto the surface of gypsum nucleus can change the surface tension of gypsum nucleus following a linear relationship proposed by Eberhart (1996) for a two components binary system (Eberhart, 1966):

211 $\sigma_{XY} = (1 - x_Y)\sigma_X + x_Y \sigma_Y$ (6)

212 where X and Y are the two components; x_Y is the mole fraction of component Y. For most cases, the threshold scale inhibitors have a small molar ratio on the surface. For example, Zieba et al. (1996) suggested that the effective active site coverage is less than 5% for phosphonate inhibitors on gypsum and barite (Zieba et al., 1996). Tomson et al. (2003) found that NTMP (nitrilomethylene phosphonic acid) occupies about 16% effective active sites of barite surface (Tomson et al., 2003; Zieba et al., 1996). Thus, the surface tension of nuclei with inhibitor adsorption can be estimated to be:

219 $\sigma_{w/\ln h} \approx \sigma_x + x_{\ln h} \sigma_{\ln h}$ (7)

220 where σ_{Inh} is the surface tension of the inhibitor; x_{Inh} is proportional to the effective coverage 221 fraction θ as $x_{\text{Inh}} = B\theta$, where *B* is the molar ratio of the active sites over mineral molecule on the nucleus surface. Thus, the surface tension of gypsum in the presence of inhibitors is:

$$
223 \t\t \sigma_{w/\text{Inh}} \approx \sigma_X + \frac{BK_L[\text{Inh}]}{1 + K_L[\text{Inh}]} \sigma_{\text{Inh}} = \sigma_X + \frac{(B\sigma_{\text{Inh}})K_L[\text{Inh}]}{1 + K_L[\text{Inh}]}.
$$
 (8)

 Considering that the total adsorbed amount of inhibitor is minimal comparing with the initial total inhibitor concentration during the induction period, the equilibrium inhibitor concentration (i.e., [*Inh*]) is estimated to be equal to the initial total inhibitor concentration. Via minimizing the 227 difference between the predicted and the measured induction time of all experiments (Table S1),

 the SI dependent interfacial energy of gypsum, effective diffusion coefficient, interfacial energy of the inhibitor, and the adsorption isotherms are optimized. The induction time of gypsum with or without the presence of inhibitors can be calculated using the new mechanistic crystallization and inhibition model by substituting Equation [\(8\)](#page-9-0) in Equation [\(2\).](#page-8-0)

3. Results and Discussions

3.1.Induction Time of Gypsum Crystallization without Inhibitors

 Based on Equation [\(2\),](#page-8-0) the surface tension of gypsum depending on temperature and gypsum *SI* is fitted to be $4.54E-4 \times (1 - 1.76E-3 \times T) \times [1 + 79.19 / (1 + \exp(-7.45 \times (SI - 0.125)))] J/m^2$. In [Figure 3](#page-10-0) is plotted the surface tension changes with *SI*, with the red box showing the gypsum *SI* range of the experimental data in this study. It is shown the gypsum surface tension is about 17.26 239 mJ/m² at large gypsum *SI* values (e.g., > 1.1). At such high *SI* values, it is suggested that nucleation is homogeneous. When *SI* value drops, the surface tension gradually decreases, with the inflection point of *SI* around 0.125. At *SI* values above 0.48, the gypsum surface tension is above 90% of the homogeneous surface tension. Such surface tension change indicates the gradual transition from pure homogenous nucleation at large *SI* values to the co-occurrence of homogenous and heterogeneous nucleation at relatively lower *SI* values. In previous research, it was proposed that mineral deposition can be initiated by two pathways: (1) homogenous nucleation in the solution followed by solids attachment to the surface, and (2) heterogeneous nucleation and crystallization directly on the surface (Huang et al., 2020; Matin et al., 2019). The second path is believed to be more energy favorable and happens at low *SI* values (Matin et al., 2019), which matches with what has been observed in this study.

251 Figure 3. The surface tension of gypsum changes with *SI* at 25 \degree C, where the red box shows the

 SI range of the experimental data. The X axis is the gypsum *SI*, and the Y axis is the surface tension of gypsum.

255 In [Figure 4](#page-11-0) is plotted the induction time of gypsum without the presence of inhibitors 256 changing with $ST²$ at different temperatures. The comparison between the measured and predicted $257 \log_{10}(t_{ind})$ at all conditions are plotted in the first subplot of [Figure 5.](#page-12-0) Note that $\log_{10}(t_{ind})$ does not 258 follow a linear relationship with ST^2 as predicted by CNT. The *SI* dependent surface tension fitted 259 above (Equation [\(3\)\)](#page-9-1) accurately explains such deviations from CNT. At higher temperatures, the 260 new model prediction shows a larger deviation, which might be due to potential phase change to 261 hemihydrate or anhydrite at higher temperatures (Blount and Dickson, 1973). It was reported that 262 the transition temperature between gypsum and anhydrite at 1 atm is around 40 \degree C and gypsum is 263 the dominant phase at temperatures below 40 °C (Lu et al., 2012; Rolnick, 1954).

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265 Figure 4. The measured (symbols) and predicted (dotted lines) base ten logarithm of the gypsum 266 crystallization induction time at different temperatures and background salt concentrations,

267 without any inhibitors.

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270 Figure 5. The predicted (Y axis) and measured (X axis) log10(*tind*) without or with nine different 271 inhibitors at different conditions. The black dotted line is the 1:1 line and the red lines are the \pm 272 0.5 error band. The linear correlation equations are also listed in each subplot.

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274 **3.2.Induction Time of Gypsum Crystallization with Inhibitors**

 In [Figure 5](#page-12-0) is plotted the comparison of the predicted and measured log10(*tind*) with the presence and absence of various inhibitors at different conditions [\(Table 2\)](#page-7-1). The background ionic strength is picked to cover the extreme conditions that may occur during various industrial processes, e.g., heat exchange, concentration polarization in membrane systems, produced water production in the oil and gas production and geotherm energy exploitation. Most of the data lie 280 along the 1:1 line within the \pm 0.5 error band, indicating the accuracy of this model. Due to the 281 relative large uncertainty of such induction time measurements, $a \pm 0.7$ error band was used in previous research (Reznik et al., 2012). In [Figure 6](#page-14-0) and [Figure](#page-15-0) 7 are plotted the specific examples showing the comparison between measured (symbols) and predicted (lines) log10(*tind*) of gypsum crystallization changing with inhibitor concentrations. It was observed that, different from that of the barite and calcite crystallization kinetics (C. Dai et al., 2019; Dai et al., 2020, 2017b; He et al., 286 1999, 1996, 1994b), the $log_{10}(t_{ind})$ does not follow a linear relationship with the inhibitor concentration, especially when the inhibitor concentrations are large. Such non-linear correlations

288 make the empirical models proposed before (i.e., $log_{10}(t_{ind}) - log_{10}(t_{ind}^0) = b \cdot C_{in}$ \Box inapplicable to the gypsum crystallization and inhibition kinetics predictions. The predicted induction times using this new model and the measured predicted induction times show good agreement, with a few exceptions. For some conditions for BHPMP, HDTMP, and PPCA, potentially due to experimental error (especially for those cases wherein the induction time is longer than 2 hours or more when large uncertainties exist) or the lack of temperature dependence of the inhibitor surface tension. Such agreement also validates the assumption that scale inhibitors prolong the induction time by adsorbing onto nucleus surface following a Langmuir type adsorption isotherm. The wide ranges of temperature, gypsum *SI* and background NaCl concentration give this model the potential to be widely applied to different industrial conditions. Furthermore once a single measurement of *tind* is made for any specific set of conditions and inhibitor concentration, calculation of the effect of changing any specific condition or concentration from the initially measured value is reliable, principally because this model is mechanistically based.

Figure 6. log10(*tind*) of gypsum crystallization under various conditions changing with the

concentrations of non-polymeric phosphonate inhibitors (i.e., BHPMP, DTPMP, HDTMP,

NTMP, EDTMP, and HEDP). The symbols are experimental measurements, and the dotted lines

- are model predictions.
-

 Figure 7. log10(*tind*) of gypsum crystallization under various conditions changing with the concentrations of polymeric inhibitors (i.e., PPCA, PVS, and CMI). The symbols are

experimental measurements, and the dotted lines are model predictions.

3.3.Adsorption of Inhibitors on Gypsum

 In [Table 3](#page-17-0) are listed the fitted *BσInh* and *K^L* values in Equation [\(8\)](#page-9-0) of different inhibitors for gypsum crystallization from this study. The *BσInh* values indicate the maximum surface tension 320 change if a large amount of scale inhibitor $(K_L[Inh] \gg 1)$ were available. For example, at large inhibitor concentrations, HDTMP and NTMP show the largest surface tension change (i.e., most 322 effective) of 117 and 120 mJ/m², respectively, while PVS shows the smallest surface tension 323 change (i.e., least effective) of 51 mJ/m². The *B* term is the molar ratio of inhibitors over the mineral molecules on the nucleus surface, which was reported to be ranging from 5% (Zieba et al., 1996, p. 199) to 16% (Tomson et al., 2003) on different minerals and inhibitor types. Thus, the superficial surface tension of each inhibitor can be estimated by dividing the *BσInh* value with the

 estimated *B* term value. The *K^L* values represent the affinity of inhibitors to the mineral nucleus. When inhibitor concentration is 1/*KL*, the surface tension change is half of the maximum surface 329 tension change (i.e., $0.5 B\sigma_{Inh}$). Thus, the inhibitors with smaller K_L values (e.g., CMI, NTMP, and HEDP) have weaker affinity to the mineral surface and require larger inhibitor concentrations (mol/L) to reach their maximum efficiency.

 The *K^L* values between the inhibitors and celestite were fitted from the celestite crystallization induction times and are listed in [Table 3](#page-17-0) (Zhao et al., 2022). It was observed that the *KL* values of the three inhibitors (i.e., DTPMP, PPCA, and PVS) with gypsum and celestite 335 follow a linear relationship with slope of 1.07 and \mathbb{R}^2 of 0.9989. That implies that the inhibitors with a strong affinity to gypsum usually represent a strong affinity to celestite as well. The *K^L* values between the inhibitors and other minerals (i.e., barite, hydroxyapatite, and calcite) were also calculated from crystal growth and dissolution rate changes. More studies to understand such differences will be valuable.

341 Table 3. The fitted *BσInh* and *K^L* values of different inhibitors for gypsum crystallization from this study, and *K^L* values for different 342 inhibitors on other minerals from literature.

 K_L (10⁶ L/mol)

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345 **3.4.Applications in Membrane System Management**

 This new model, by substituting Equation [\(8\)](#page-9-0) in Equation [\(2\),](#page-8-0) can be used in understanding the gypsum crystallization and deposition (or fouling). For example, Huang et al. (2020) studied the impacts of surface hydrophobicity on gypsum deposition (Huang et al., 2020). They ran the testing at two different conditions [\(Table 4\)](#page-18-0) to represent the homogeneous nucleation pathway at a higher *SI* (Condition 1) and the heterogeneous nucleation pathway at a smaller *SI* (Condition 2), respectively. Under Condition 1, the predicted induction time and the two measured induction times are probably within experimental error. In addition, under this condition, the surface tension is more than 90% of the homogenous nucleation surface tension (Equation [\(3\)\)](#page-9-1), indicating that homogeneous nucleation will be dominated, which matched reasonably well with their experimental results. Under Condition 2, the predicted induction time matched well with the induction times measured by He et al. (1996) and Huang et al. (2020). Different than Condition 1, Huang et al. (2020) believed that under Condition 2 heterogeneous nucleation is dominated and thus they used QCM-D (quartz crystal microbalance with dissipation) to monitor the deposition onto membrane material surfaces, instead of using DLS (dynamic light scattering) to monitor bulk crystallization. According to the mineral crystallization surface deposition platform our group developed (Dai et al., 2021; Lu et al., 2020a, 2020b, 2019), it is believed that the first phase deposition happens in the stagnant laminar sublayer following the same bulk crystallization kinetics. The good agreement between the detection of surface deposition by Huang et al. (2020) with the bulk crystallization induction time measured by He et al. (1994a) and predicted by this study further confirms the validity of the gypsum crystallization model proposed in this study and the mineral crystallization surface deposition platform proposed previously (Dai et al., 2021). 367

368 Table 4. Two gypsum crystallization conditions in Huang et al. (2020) and the comparison of the 369 induction times by this study, He et al. (1994a) and Huang et al. (2020).

Experimental Conditions	$log_{10}(t_{ind})$		
	This	He et al.	Signal change in
	study	(1994a)	Huang et al. (2020)
$\overline{1: Na^+}$, Cl ⁻ , Ca ²⁺ , and SO ₄ ²⁻ concentrations		2.70	$3.32(35 \text{ min})$
are 0.1, 0.1, 0.05, and 0.05 mol/L, $T = 22$ °C,	2.50	$(SI = 0.50,$	(DLS)
$SI = 0.51$ (SSP) and 0.52 (Visual Minteq)		$T = 25$ °C)	
2: Na^+ , Cl ⁻ , Ca^{2+} , and $SO_4{}^{2-}$ concentrations are 0.05, 0.05, 0.025, and 0.025 mol/L, $T =$ 22 °C, $SI = 0.15$ (SSP) and 0.15 (Visual Minteg)	4.52	4.38 $(SI = 0.15,$ $T = 25 \text{ °C})$	$4.55 \left(\sim 10 \text{ hours} \right)$ $(QCM-D)$

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371 The utility of having this complete crystallization and inhibition model is apparent: in 372 addition to predicting the initial induction times, the impact on crystallization by varying the 373 conditions (temperature, saturation index, etc.) or addition of small amounts of threshold inhibitor can readily be calculated. For example, under Condition 1, the addition of 1 mg/L of PPCA can 375 prolong the induction time to about 17.5 hours (i.e., $10^{4.80}$ seconds). That means, in a membrane water treatment system, the addition of 1 mg/L of PPCA in the concentrate side can significantly reduce the periodical feed water flush frequency by a factor of 30 times if keeping the same recovery ratio.

4. Conclusions

 The unexpected mineral scale deposition and fouling in the membrane system has been one of the main reasons that limit the wide applications of membrane treatment technology. However, few mechanistic models are available to accurately predict the induction time of gypsum, one of the most common and troublesome scale minerals. This study developed a new crystallization and inhibition model based on the classical nucleation theory. This model assumes that the gypsum surface tension is dependent on *SI*. At more supersaturated conditions, the surface tension is larger, and is homogeneous nucleation dominated; at less saturated conditions, the surface tension gradually decreases with more occurrences of heterogeneous nucleation. A sigmoid function is used to describe such gradual transition from homogenous to heterogeneous nucleation. This model accurately predicts the gypsum crystallization induction time at conditions of *SI* from 0.04 391 to 0.96, temperature from 25 to 90 °C, background NaCl salt concentration from 0 to 6 mol/L.

 A Langmuir-type adsorption isotherm is adopted to describe the adsorption of inhibitors to the gypsum nucleus surface. The surface tension of the gypsum nucleus will increase due to the inhibitor adsorption. This study used this new model and accurately predicted the induction time of gypsum crystallization in the presence and absence of nine commonly used scale inhibitors. The fitted affinity constants (*KL*) between these inhibitors and gypsum shows a good linear relationship with those between the same inhibitors and barite, implying a similar adsorption mechanism.

 This study successfully adopts the new crystallization model and resolves the long-unsolved problem of gypsum crystallization kinetics prediction with or without the dosage of various scale inhibitors. By incorporating the mineral surface deposition model developed before, this model can accurately predict the deposition time of gypsum on the membrane material surface. The utility of this model can help accurately evaluate the impacts of temperature, saturation index, and scale inhibitors. Such predictions can help improve the operational efficiency of membrane systems, including product water recovery, periodical feed water flush frequency, scale inhibitor type and dosages at different operating conditions, to mention a few.

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