Research Article

Reaction Kinetics of Cinnamaldehyde Hydrogenation over Pt/SiO₂: Comparison between Bulk and Intraparticle Diffusion Models


1 Chemical Engineering Department, University of Technology-Iraq, Baghdad 10066, Iraq
2 Chemical Engineering and Oil Refinery Department, AlMustaqbal University College, Hilla, Babylon, Iraq
3 Engineering College, Misan University, Misan, Iraq
4 University of Sheffield, Department of Chemical and Biological Engineering, Mappin Street, Sheffield S1 3JD, UK

Correspondence should be addressed to Adnan A. AbdulRazak; adnansss2002@yahoo.com

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The liquid-phase hydrogenation of cinnamaldehyde over a Pt/SiO₂ catalyst was investigated experimentally and theoretically. The experiments were conducted in a 300 cm³ stainless steel stirred batch reactor supplied with hydrogen gas and ethanol as a solvent. Five Langmuir–Hinshelwood kinetic models were investigated to fit the experimental data. The predictions from the bulk model were compared with predictions from the intraparticle diffusion model. Competitive and non-competitive mechanisms were applied to produce the main intermediate compound, cinnamyl alcohol. Reaction rate parameters for the different reaction steps were calculated by comparing between the experimental and mathematical models. All rate data utilized in the present study were obtained in the kinetic regime. The kinetic parameters were obtained by applying a nonlinear dynamic optimization algorithm. Nevertheless, the comparison between the methodology of the present model and these five models indicated that the non-competitive mechanism is more acceptable and identical with the single-site Langmuir-Hinshelwood kinetic model including mass transfer effects and it mimicked the reactant behavior better than the other models. In addition, the observed mean absolute error (MAE) for the non-competitive mechanism of the present model was 2.3022 mol/m³; however, the MAE for the competitive mechanism was 2.8233 mol/m³, which is an increase of approximately 18%. The prediction of the intraparticle diffusion model was found to be very close to that of the bulk model owing to the use of a catalyst with a very small particle size (<40 microns). Employing a commercial 5% Pt/SiO₂ catalyst showed a result consistent with previous research using different catalysts, with an activation energy of ≈24 kJ/mol.

1. Introduction

During recent decades, increasing global population has fueled demand for energy vectors, and commodity, specialty, and fine chemicals. One important group of chemicals which has seen such growth in demand is α,β-unsaturated aldehydes. This has forced a reevaluation of the current methods used to produce these substances [1–3]. Of particular note is cinnamaldehyde (CALD), the hydrogenation of which in the liquid phase generates valuable chemicals through consecutive reactions [4]. This reaction provides an excellent opportunity to study the influence of catalyst and operating conditions on the reactant conversion and selectivity in molecules containing multiple reactive functionalities and serves as an exemplar molecular for other α, β-unsaturated aldehydes and ketones.

Cinnamaldehyde consists of conjugated C=C and C=O bonds connected to an aromatic ring [1, 5, 6]. Borovinskaya [7] and Chang et al. [4] studied the reaction pathway of the direct hydrogenation of cinnamaldehyde to hydrocinnamal alcohol; however, other researchers have identified two indirect pathways via cinnamyl alcohol of hydrocinnamaldehyde, as shown in Figure 1 [1, 2, 5, 8–24].
Cinnamaldehyde
(CALD)

Cinnamyl alcohol
(CALC)

Hydrocinnamaldehyde
(HALD)

Hydrocinnamyl alcohol
(HALC)

Figure 1: Reaction scheme of cinnamaldehyde hydrogenation.

Of the three possible hydrogenation products, cinnamyl alcohol is the most favored as it has a wide variety of applications, including as a flavor in bakery goods, confec-

tionary, and beverages; as an animal or flea repellent; as protection for plants against nematodes; as an antimicrobial agent; as a component of air fresheners; and in the medical

field [4, 25]. In general, C=C hydrogenation is thermody-

namically favored whereas alcohol desorption is slower which means that hydrocinnamaldehyde is often the main product. In this reaction, the production of cinnamyl alcohol as an intermediary compound is considered the major challenge. However, choosing a catalyst with a suitable structure and composition can overcome this challenge and increase the production of cinnamyl alcohol. The type of catalyst plays an important role in the present reaction because they can increase the selectivity of cinnamyl alcohol (CALC). Important factors affecting activity and selectivity include the active metal, the properties of the catalyst support, and the particle size of the metal. Pt/SiO$_2$ was a selective catalyst of the series toward cinnamyl alcohol; achieving high selectivity to hydrogenation of the C=O functionality over the active sites is challenging. While Pt catalysts favor this over hydrogenation of the C=C functionality, subsequent hydrogenation results in the production of hydrocinnamyl alcohol as a byproduct [1–3].

The selectivity of cinnamaldehyde hydrogenation to cinnamyl alcohol (CALC) has been extensively studied [1–23]; however only a few kinetic studies have investigated the reaction kinetics of cinnamaldehyde hydrogenation.

The most common model proposed to describe this reaction mechanism is the Langmuir–Hinshelwood model (LHM) (or Langmuir–Hinshelwood–Hougen–Watson (LHHW) model). For example, Mohire and Yadav [26] studied the hydrogenation of cinnamaldehyde to hydro-

cinnamaldehyde over a 5% Ni–Cu catalyst and assumed that cinnamaldehyde and H$_2$ both adsorbed on the vacant surface sites. By calculating the reaction rate constants at different temperatures, the activation energy was observed to be 52.05 kJ/mol. Elsewhere, Hajek and Murzin [27] calculated an activation energy of ~34 kJ/mol for cinnamaldehyde hydrogenation over a Ru-5% Sn/SiO$_2$ catalyst. Although a

plausible description of the concentration dependencies was obtained via fitting to the Langmuir–Hinshelwood model, the model failed to fit the selectivity dependence.

In contrast, Khan et al. [1] and Neri et al. [28] described the effect of various operating parameters on the reaction kinetics of liquid-phase cinnamaldehyde hydrogenation over both a 5% Pt/C and a Ru/Al$_2$O$_3$ catalyst. The authors employed the LHM; however they assumed that both the adsorption and hydrogenation of C=O and C=C groups occur at different active sites with the competitive adsorption of reactants and products. Adsorption and hydrogenation of the C=O bond increases the selectivity to cinnamyl alcohol.

Yamada et al. [29] studied the reaction rate for the hy-

drogenation of cinnamaldehyde over Pd/C and Pt/C catalysts using aqueous potassium hydroxide solution as a promoter. The reaction rate was observed to be of zero-order with respect to the unsaturated aldehydes. Liu et al. [30] achieved high conversions (>99.9%) and high selectivity (>99.9%) when employing an AlCl$_3$–Pd/C dual catalyst at ambient temperature and atmospheric pressure. The estimated Langmuir–Hinshelwood kinetic model indicated that the reaction was zero-order in cinnamaldehyde, whereas it was first-order in hydrogen.

The reaction rate is controlled by the mass transfer of the reactant from the bulk liquid phase to the surface of the catalyst particles if external liquid-solid mass transfer limit-

ations exist [31]. Toebes et al. [32] used a single-site model incorporating both Langmuir–Hinshelwood kinetics and mass transfer to describe the hydrogenation of cinnamalde-

hyde over a carbon nanofiber-platinum catalyst. They de-

ected that the intrinsic reaction rate grew by a factor of 120 with the removal of the oxygen-containing surface groups.

In summary, the literature reveals that different kinetic models have been used to describe the consecutive hydrogenation of cinnamaldehyde; however, these kinetic models are diverse in mechanism, and hence in the accuracy of their representation of the experimental results. Therefore, the present work aims to mathematically model the reaction kinetics of cinnamaldehyde hydrogenation over a commercial 5% Pt/SiO$_2$ catalyst. The main objective of this study is to increase the production of cinnamyl alcohol, which appears as an intermediate product with hydrocinnamaldehyde, by hinder-

ing the hydrogenation reactions that lead to its conversion to hydrocinnamyl alcohol. In addition, this research will predict the kinetic parameters of the Langmuir–Hinshelwood model by employing two mathematical models: bulk and intraparticle diffusion. The proposed mathematical model deals with two types of active sites, namely, the acidic sites and the metallic sites represented by the platinum loaded on the surface of the SiO$_2$ catalyst. The model in this study suggests that there is an abundance of hydrogen gas and therefore the competition between hydrogenation/dehydrogenation reactions taking place on active sites and on the other hand the state of non-

competition has been taken into account.

2. Experimental Work

Experiments were conducted using a 300 cm$^3$ stainless steel batch reactor (Parr Instrument Company, USA) equipped with a pressure gauge, stirrer, and heater to control the
temperature, which was monitored via a thermocouple, as shown in Figure 2. A commercial 5% Pt/SiO₂ (Esctat 2351) catalyst supplied by Strem Chemicals (USA) was employed directly without further modification (gray powder, particle size = 40 μm, surface area = 400 m²/g). Ethanol (Sigma-Aldrich, USA) was used as the solvent. For a typical experiment, 0.1 g of catalyst, 10 g of cinnamaldehyde (99% GC purity, purchased from Sigma-Aldrich), and 190 mL of solvent were introduced into the reactor. The reactor was then flushed three times with 5 bar nitrogen; the temperature was increased moderately to the required value (90–110°C) using heating oil; and the pressure of hydrogen was raised to the desired value (10–30 bar). A reactor was equipped with a pressure control valve which was used to control the reactor pressure by regulating the hydrogen being fed to the reactor until the desired operating conditions were reached. Mass transfer can play an important role in liquid-phase hydrogenations using porous catalysts. Therefore, it is critically important that mass transfer effects are ruled out before attempting to obtain reliable kinetic data. If external liquid-solid mass transfer limitation is negligible and the internal diffusion effects are significant, then the concentration gradient across the catalyst pellet must be taken into consideration in determining the net reaction rate. A rigorous mathematical model was developed for a diffusion-reaction process in a spherical catalyst pellet contained in an isothermal batch reactor.

3. Mathematical Modelling

3.1. Bulk Model. Bulk models assumed that (i) each point on the interior of the catalyst surface is accessible to the same concentration [33] and (ii) mixing provides a uniform distribution of liquid and catalyst particles. The adsorption reaction kinetic models suggest that the adsorption rate of the solute on the surface of the adsorbent has the dominant influence on the adsorption kinetics; therefore, the exterior mass transfer and intraparticle diffusion can be considered negligible [34].

For the reaction pathway shown in Figure 1, the set of differential equations (1) to (4) expresses the concentration change in the bulk liquid for the four components [27]:

\[
\frac{dc_{\text{CALD}}}{dt} = \frac{W_c}{V_R} (r_1 + r_2), \quad (1)
\]

\[
\frac{dc_{\text{CALC}}}{dt} = \frac{W_c}{V_R} (r_1 - r_3), \quad (2)
\]

\[
\frac{dc_{\text{HALD}}}{dt} = \frac{W_c}{V_R} (r_2 - r_4), \quad (3)
\]

\[
\frac{dc_{\text{HALC}}}{dt} = \frac{W_c}{V_R} (r_3 - r_4). \quad (4)
\]

3.2. Intraparticle Diffusion Model. Two steps summarize the intraparticle diffusion model, including diffusion of the reactant molecules: firstly at the bulk external film, and secondly between the exterior film and the surface of catalyst, where the reaction occurs [35]. If external liquid-solid mass transfer limitation is negligible and the internal diffusion effects are significant, then the concentration gradient across the catalyst pellet must be taken into consideration in determining the net reaction rate. A rigorous mathematical model was developed for a diffusion-reaction process in a spherical catalyst pellet contained in an isothermal batch reactor.

3.2.1. Mass Balance in the Bulk Liquid. The mass balance of the bulk liquid is expressed below [32]:

![Diagram](image_url)
\[
\frac{dc_{b,j}}{dt} = \frac{A_p}{V_{liq}} D \frac{\partial c_{b,j}}{\partial r} |_{r=r_p},
\]
\[
A_p = \frac{3}{r_p} V_p.
\]

3.2.2. Mass Balance in the Catalyst Particle. The mass balance inside the catalyst particle is given by [36-38]

\[
\frac{dc_{p,j}}{dt} = D_{eff,j} \left( \frac{c_{p,j,i+1} - 2c_{p,j,i} + c_{p,j,i-1}}{\Delta r^2} + \frac{2 c_{p,j,i+1} - c_{p,j,i-1}}{2\Delta r} \right) + (1 - \varepsilon_p)v_j \rho_{cat} R_{r,j},
\]

Equation (7) has been discretized by a finite difference method based on a three-point centered scheme for the second-order derivative.

\[
\frac{dc_{p,j}}{dt} = D_{eff} \left( \frac{c_{p,j,i+1} - 2c_{p,j,i} + c_{p,j,i-1}}{\Delta r^2} + \frac{2 c_{p,j,i+1} - c_{p,j,i-1}}{2\Delta r} \right) + (1 - \varepsilon_p)v_j \rho_{cat} R_{r,j},
\]

with boundary conditions

\[
\begin{align*}
\varepsilon_p &= 0, \\
\frac{dc_{p,j}}{dr} &= 0, \\
\varepsilon_p &= r_p, \\
c_{p,j,i} &= c_{p,j,i} (r = r_p).
\end{align*}
\]

where \( j \) represents the component index and \( i \) represents the radial position inside the reactor particles. Equation (8) has been generalized for five components within the reaction mixture.

The reactant concentrations around the catalyst particle is assumed to be equal to the reactant concentrations in the bulk. The Wilke–Chang equation (equation (9)) was used to determine the diffusion coefficient for a dilute solute \( j \) [39, 40]:

\[
D^0_{jm} = 7.4 \times 10^{-8} \sum_{j=1}^{N_A} \left( x_j \mu_j M_w^j \right)^{0.5} T
\]

For the multicomponent liquid mixture, the diffusivity of compound \( j \) was estimated, as per equation (10) [41]:

\[
D_{jm} = \frac{D^0_{jm}}{r}.
\]

Rigorous mathematical modelling involves solving the differential equations of the reactant concentrations in bulk liquid and inside catalyst particle.

4. Reaction Kinetic Models

Two main reasons make employing a batch reactor better than a continuous reactor for reaction kinetic studies. First, in a batch reactor, the changes in the reactant concentrations can be measured instantaneously, whereas the continuous reactor provides only the final concentrations. Second, in a batch reactor, the reaction temperature is controlled accurately, allowing for a precise description of the reaction rate constant with respect to temperature. Selecting an accurate kinetic model is essential for the optimal representation of the experimental results and thus for a better design that optimizes the process. In addition, accurate kinetic models can maximize the selectivity of the desired products. Although many researchers have investigated cinnamaldehyde hydrogenation over different types of catalysts using the Langmuir–Hinshelwood model to fit the experimental data, no comparative study has distinguished between these kinetic models. Zamostny and Belohlav [42] used a simplified Langmuir–Hinshelwood kinetic model to fit experimental data of cinnamaldehyde hydrogenation. Hydrogen pressure was not taken into consideration in the kinetic modelling and the adsorption coefficient was assumed to be independent of temperature. Hajek and Murzin [27] employed a Langmuir–Hinshelwood reaction mechanism with the assumption of different adsorption sites for organic compounds and hydrogen. Virtanen et al. [43] used the Langmuir–Hinshelwood–Hougen–Watson concept to represent cinnamaldehyde hydrogenation, assuming competitive adsorption of all compounds with dissociative hydrogen adsorption. Single-site Langmuir–Hinshelwood kinetic model including mass transfer effects was developed by Toebes et al. [32]. It was assumed that no competition takes place with hydrogen adsorption, which is assumed to occur dissociatively. A two-site Langmuir–Hinshelwood kinetic model was developed by Khan et al. [1]. This model considered the effect of various operating parameters (catalyst loading, partial pressure of hydrogen, and initial substrate concentration) on the rates of hydrogenation. A summary of previously investigated kinetic models is presented in Table 1 in order to facilitate comparison.

The Arrhenius equation was employed to elucidate the relationship between the reaction rate constant \( k \) and the reaction temperature.

\[
k_i = A_i \exp \left( -\frac{E_i}{RT} \right).
\]

It was assumed that the adsorption equilibrium constants do not depend on temperature.

As seen in Figure 1, the reaction kinetic model of the present study (RKP) was investigated by employing Langmuir–Hinshelwood kinetics. The goal was to reach the optimal reaction mechanism that results in the largest production of cinnamyl alcohol. However, throughout the experiments, small amounts of hydrocinnamaldehyde (HCNAL) were detected in the bulk solvent. Reaction modelling requires that the reaction pathway of cinnamaldehyde to hydrocinnamaldehyde, as well as hydrocinamyl alcohol, be included in the model.

The model assumed that the adsorption of reactants and products is reversible and competitive while hydrogen is
activated on different sites and does not compete with the organic compounds (hydrogen shows non-competitive and dissociative adsorption). So adsorption and desorption are considered to be in quasi-equilibrium. The rate-determining step is considered to be the reaction between atomic hydrogen and the adsorbed organic compound. Therefore, a one site model was investigated. In this model, C=C and C=O are hydrogenated on the same type of site, as shown below:

\[ r_1 = \frac{K_1 K_A C_A}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

\[ r_2 = \frac{K_2 K_A C_A}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]  

\[ r_3 = \frac{K_3 K_B C_B}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

A: cinnamaldehyde, B: cinnamyl alcohol, C: hydrocinnamaldehyde, D: hydrocinnamal alcohol.

Because the present study was conducted at high pressure, the hydrogen term is neglected because the order associated with hydrogen is zero, so the value of the numerator is equal to the value of the denominator. Therefore, the limit of hydrogen will be one, as shown in the following:

\[ r_4 = \frac{K_4 K_C C_C}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

\[ r_1 = \frac{K_1 K_A C_A}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

\[ r_2 = \frac{K_2 K_A C_A}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

\[ r_3 = \frac{K_3 K_B C_B}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]

\[ r_4 = \frac{K_4 K_C C_C}{1 + K_A C_A + K_B C_B + K_{C=C} C_C + K_D C_D} \times \frac{\sqrt{K_H C_{H_2}}}{1 + \sqrt{K_H C_{H_2}}} \]  

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5. Computational Method

The kinetic parameters were estimated from experimental data by means of a nonlinear dynamic optimization algorithm, in which results of the mathematical model were compared against the experimental information. Six experiments at different operating conditions were simulated for each reaction mechanism, and the mean absolute error (MAE) was estimated by comparing the simulation concentrations with the predicted experimental concentrations. The MAE was calculated according to equation (20) [44, 45].

\[
\text{MAE} = \frac{1}{N \times M} \sum_{j=1}^{M} \sum_{i=1}^{N} |c_{i,j}^{\text{exp}} - c_{i,j}^{\text{pred}}|.
\]  

The ordinary differential equations governing the reactor and kinetic models were solved using the fourth-order Runge–Kutta method, while a Genetic Algorithm (GA) optimization stochastic technique was used to predict the kinetic parameters by minimizing the objective function (minimum MAE). The simulation results that were obtained from the constructed kinetic models were compared with relevant experimental results, until the model was capable of reaching reasonable results.

The present model suggests that the use of Pt as a metal support through the hydrogenation of the reaction of cinnamaldehyde contributes to an increase in the production of cinnamyl alcohol and makes it the main product. The addition of Pt reduces the free active sites of the catalyst by poisoning some or most of them according to the loading condition of Pt reduces the free active sites of the catalyst by poisoning some or most of them according to the loading.

Genetic Algorithms are powerful and widely applicable stochastic search and optimization methods based on the concepts of natural selection and natural evaluation [46]. Genetic Algorithms apply the principles of survival of the fittest, selection, reproduction, crossover, and mutation of individuals to obtain a new and better individual (i.e., new solutions). Crossover is implemented by selecting a random point on the chromosome where the parents’ parts exchange happens. The crossover then brings up a new offspring based on the exchange point chosen with particular parts of the parents. Normally, mutation takes place after crossover is done. This operator applies the changes randomly to one or more “genes” to produce a new offspring, so it creates new adaptive solutions which avoid local optima. Within the selection stage, individual genomes are chosen from a population for later breeding [47].

A nonlinear constraint was imposed on the variable’s candidate for optimization to limit the GA search span to gain acceptable kinetic parameters. Table 2 shows the parameters used in the Genetic Algorithm optimization. All computations within this study were performed using MATLAB software version 2015a.

### Table 2: Genetic algorithm parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Population size</td>
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</tr>
<tr>
<td>Maximum generation</td>
<td>1000</td>
</tr>
<tr>
<td>Crossover probability</td>
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<tr>
<td>Mutation probability</td>
<td>0.4</td>
</tr>
<tr>
<td>Neighborhood size</td>
<td>0.05</td>
</tr>
</tbody>
</table>

6. Results and Discussion

6.1. Comparison between Kinetic Models. Nonlinear optimization was used to predict the kinetic parameters of five models and compared with the methodology used in the present study. This method covered the change in the initial cinnamaldehyde concentration, temperature, and pressure on the rate of reaction for all experiments simultaneously. In fact, nonlinear optimization is more accurate than another traditional method which estimates the kinetics parameters via the linearized Arrhenius equation (ln k versus 1/T). Figure 3 illustrates the priority curves obtained for the model of the present study and five Langmuir–Hinshelwood kinetic models, where the points clustered around the diagonal line indicate the validity of these five kinetic models. Figures 3(a) and 3(g) show a divergence between the experimental and predicted results, especially for CALD; however, the other figures demonstrate an acceptable convergence. The values of the predicted errors for the present modified model and the five tested reaction kinetics are summarized in Table 3. In general, the accuracy of the obtained results increases with an increasing number of factors in the empirical equation, in which the kinetic models RKP, RK2, RK3, and RK4 have more factors (e.g., adsorption factors) than RK1 and RK5; therefore, the estimated error (MAE) for these three models (≈2.3, 2.3, 2.5 and 2.3 mol/m³, respectively) is lower than for the other two models (≈3.1 and 3.6 mol/m³, respectively). The MAE obtained by the RKP, RK2, and RK4 kinetic models were approximately equal and were smaller than that obtained by the other three kinetic models. Indeed, RK2 and RK4 have approximately the same main factors, except for the term that represents hydrogen pressure, which is equal to 1 for RK2 but is 0.5 for RK4. The estimated error for RK4 was slightly lower than for RK2 and identical to RKP, and this supports the assumption that there is a single-site for the adsorption of hydrogen molecules in the kinetic model and RKP compared with the dual-site adsorption for hydrogen molecules in the RK2 kinetic model.

When comparing the RKP model in the present study, which is directed towards the production of cinnamyl alcohol, it is noticed that there is a difference from the reaction mechanism of the RK3 model, which is directed towards the production of hydrocinnamaldehyde. Therefore, the results of the RKP model suggest that the non-competitive hypothesis for components is better than the competitive hypothesis, while the RK3 model showed that the competitive hypothesis is better. On the other hand, RK2 and RK4 models only used the non-competitive hypothesis and neglected the competitive hypothesis. The present study, which used both hypotheses, demonstrated a consistency with the trend of RK2 and RK4 models by favoring the non-competitive hypothesis over the competitive hypothesis. Furthermore, the MAE results of the RKP model in the competitive and non-competitive hypotheses are 2.8233 and 2.3022 mol/m³, respectively; these reinforce the fact that the non-competitive hypothesis is preferable. In other words, the error has been reduced by about 18%.
Figure 3: Continued.
ratio and due to the formation of hydrocinnamyl alcohols occurring at the free catalytic sites. Therefore, the selectivity of cinnamyl alcohol will increase and the hydrogenation of cinnamyl alcohol will decrease. Assuming different adsorption sites for organic compounds and hydrogen, the role of hydrogen adsorption in hydrogenation processes is interesting. On the one hand, it is known that hydrogen absorbs on noble metals; on the other hand, hydrogen molecules are much smaller than organic molecules, especially in the present case. This means that the interstitial sites between the adsorbed organic molecules remain available for hydrogen adsorption. Thus, the adsorption behavior turns into a non-competitive one. Therefore, the RK4 model was found to be better than the other four models in representing the consecutive hydrogenation of cinnamaldehyde to other products.

6.2. Comparison between Bulk and Intraparticle Diffusion Models. Cinnamaldehyde hydrogenation was carried out at temperature range 90–110°C, with a pressure range between 10 and 30 bar, using ethanol as a nonpolar solvent. Cinnamyl alcohol, hydrocinnamaldehyde, and hydrocinnamyl alcohol are the principal products that result from this reaction. Figure 4 compares the experimental results and the bulk model results using the optimum kinetic model (RKP), while Figure 5 presents a comparison between the experimental results and intraparticle diffusion model results using the same model. Both figures show a good agreement between the concentrations obtained in the experimental study and those predicted by the theoretical model.

Figures 4 and 5 provide an indication of the behavior that occurs in the hydrogenation of cinnamaldehyde, which appears to follow the behavior of a sequential parallel reaction system. Furthermore, the concentration of cinnamyl alcohol rose rapidly at the start of the reaction and then became almost constant after 200 min, similar to previous results [8]. The concentration of hydrocinnamaldehyde however slightly decreased after 50 min of reaction time. Hajek and Murzin [27] observed the same behavior, where the selectivity of hydrocinnamaldehyde decreased slightly over time.

Figure 6 presents a comparison between the estimated molar fractions obtained by the bulk and intraparticle diffusion models versus the experimental molar fractions for all experiments and components. There was good conformity between the experimental and theoretical results, as illustrated in the value of the absolute error that was determined according to the consistency between the practical and predicted mole fractions. In spite of employing the same kinetic model (RK present) for both bulk and intraparticle diffusion models, the bulk model showed a slightly low level of error (2.3 mol/m³) compared with the intraparticle model (2.33 mol/m³). This can be attributed to the error corresponding to the intraparticle mass transfer coefficient calculations.

The kinetic parameters of the reaction (RKP) for both the bulk and intraparticle diffusion models were determined by benchmarking against the experimental results. Table 4 shows the summary of the estimated errors for different kinetic models. The RKP model was found to be the best among the models tested.
displays the estimated values of the pre-exponential factor and apparent activation energy ($E_a$). Despite there being a considerable difference between the kinetic parameters calculated using the two different mathematical models, the observed MAE for both models was very similar. This can be attributed to using fine catalyst particles (40 $\mu$m), which significantly reduces the influence of mass transfer effects; therefore the variation in concentration through the catalyst display the estimated values of the pre-exponential factor and apparent activation energy ($E_a$). Despite there being a considerable difference between the kinetic parameters calculated using the two different mathematical models, the observed MAE for both models was very similar. This can be attributed to using fine catalyst particles (40 $\mu$m), which significantly reduces the influence of mass transfer effects; therefore the variation in concentration through the catalyst
particles is small. This conclusion is supported by the work of Zhu and Zaera [5], who found that the elementary hydrogenation selectivity was independent of the particle size, especially when the size was less than 250 µm. The values of the activation energy for cinnamaldehyde hydrogenation to cinnamyl alcohol were 24.33 and 31.35 kJ/mol for the bulk
and intraparticle diffusion models, respectively. These results approached the value of the activation energy (31.4 kJ/mol) obtained by Hajek and Murzin [27] for cinnamaldehyde hydrogenation employing a Ru-5% Sn/Sol-Gel catalyst, and by Mohire and Yadav [26], who obtained an activation energy of approximately 52 kJ/mol for the hydrogenation of cinnamaldehyde over a 5% Ni–Cu catalyst. From inspection of Table 4 and Figures 4 and 5, it is seen that the product in the greatest concentration was cinnamyl alcohol. Furthermore, there was a good agreement between the estimated and experimental concentrations. The activation energy of cinnamyl alcohol hydrogenation (≈152 kJ/mol) was significantly higher than that of cinnamaldehyde reduction (≈24 kJ/mol). This therefore limits the further hydrogenation of cinnamyl alcohol to form hydrocinnamyl alcohol. This result is similar to that obtained by Khan et al. [1].

As shown in Figure 1, there are two reaction routes to obtain hydrocinnamyl alcohol. Either the cinnamaldehyde hydrogenates to cinnamyl alcohol, and then to hydrocinnamyl alcohol; or, the cinnamaldehyde hydrogenates to hydrocinnamaldehyde, and then to hydrocinnamyl alcohol. If internal mass transfer limitations exist, the observed reaction rate is influenced by the diffusion of the reactants (cinnamaldehyde) from the surface to the active sites and reverse diffusion of intermediate and final products (cinnamyl alcohol, hydrocinnamaldehyde, and hydrocinnamyl alcohol) from active sites to the catalyst surface. Therefore, the activation energies for cinnamaldehyde reduction and cinnamaldehyde hydrogenation in the intraparticle diffusion model were higher than the corresponding reactions in the bulk model. In contrast, the activation energies for the cinnamyl alcohol hydrogenation and hydrocinnamaldehyde reduction in the intraparticle diffusion model were lower than the corresponding reactions in the bulk model. This occurred because in the intraparticle model, the two reactions (cinnamaldehyde reduction and cinnamaldehyde hydrogenation) occur after cinnamaldehyde diffuses into the catalyst particles. After this, the intermediate products react, and then the reaction products are liberated from the catalyst as a result of diffusion.

### 7. Conclusion

The reaction kinetics of the liquid-phase hydrogenation of cinnamaldehyde over a Pt/SiO₂ catalyst were investigated. The modified model and the five kinetic models for cinnamaldehyde hydrogenation were evaluated and compared. The Langmuir–Hinshelwood kinetic model, which assumes a single-site adsorption for hydrogen molecules, was found to best represent the consecutive hydrogenation of cinnamaldehyde. In addition, the prediction of the bulk diffusion

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**Table 4: Predicted kinetic parameters.**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Bulk model</th>
<th>Intraparticle diffusion model</th>
</tr>
</thead>
<tbody>
<tr>
<td>Adsorption constants</td>
<td>K_{CALD}</td>
<td>0.0298</td>
</tr>
<tr>
<td></td>
<td>K_{CALC}</td>
<td>0.9514</td>
</tr>
<tr>
<td></td>
<td>K_{HALD}</td>
<td>0.1663</td>
</tr>
<tr>
<td></td>
<td>K_{HALC}</td>
<td>0.9764</td>
</tr>
<tr>
<td></td>
<td>K_{H₂}</td>
<td>0.0052</td>
</tr>
<tr>
<td>Pre-exponential factor</td>
<td>A₁₁₀</td>
<td>64.1472</td>
</tr>
<tr>
<td></td>
<td>A₂₁₀</td>
<td>98.3952</td>
</tr>
<tr>
<td></td>
<td>A₃₁₀</td>
<td>534.75</td>
</tr>
<tr>
<td></td>
<td>A₄₁₀</td>
<td>199.63</td>
</tr>
<tr>
<td>Activation energy (kJ/mol)</td>
<td>E₁</td>
<td>24.33</td>
</tr>
<tr>
<td></td>
<td>E₂</td>
<td>28.27</td>
</tr>
<tr>
<td></td>
<td>E₃</td>
<td>152.59</td>
</tr>
<tr>
<td></td>
<td>E₄</td>
<td>37.74</td>
</tr>
</tbody>
</table>

---

**Figure 6:** Parity diagram indicating the calculated molar fractions obtained by the (a) bulk and (b) intraparticle diffusion models versus the experimental molar fractions for all experiments and components.
model was compared with the intraparticle diffusion model. 
The predictions of the intraparticle model (MAE = 2.33 mol/
per m³) were very close to those of the bulk model
(MAE = 2.30 mol/m³) due to using a catalyst with a very
small particle size (< 40 microns). Although the MAE results
are convergent, the bulk model is more favorable as it is
simpler than the intraparticle model.

Consequently, the present model and the other five
kinetic models for cinnamaldehyde hydrogenation were
evaluated and compared. The Langmuir–Hinshelwood ki-
netic model, which assumes a single-site adsorption for
hydrogen molecules, was found to best represent the con-
ssecutive hydrogenation of cinnamaldehyde. The bulk model
is more favorable as it is simpler than the intraparticle
model. The study concluded that the non-competitive
mechanism between unsaturated hydrogenation reactions to
produce intermediate compounds, including cinnamyl al-
cohol, and saturated hydrogenation reactions to increase the
production of either the intermediate compound (i.e.,
hydrocinnamaldehyde) or the hydrogenation of the inter-
mediate compounds and its complete conversion to
hydrocinnamyl alcohol is the most reasonable compared to
the case of competitive, which has been proposed in other
studies, as there is no competition taking place between
disruptive reactions and cinnamaldehyde hydrogenation.

The activation energies of cinnamaldehyde hydrogena-
tion to cinnamalcohol were ~31 kJ/mol and ~24 kJ/mol for
the intraparticle and the bulk models, respectively.
Employing a commercial 5% Pt/SiO₂ catalyst showed a result
consistent with previous reports that employed different
types of catalysts, although with a lower activation energy of
~24 kJ/mol.

**Nomenclature**

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>(A_o)</td>
<td>External area between the bulk fluid and the particle (m²/s)</td>
</tr>
<tr>
<td>(A_p)</td>
<td>Pre-exponential factor (mol/gm.cat.min)</td>
</tr>
<tr>
<td>(c_j)</td>
<td>Concentration of component (j) (mol/m³)</td>
</tr>
<tr>
<td>(c_{vb})</td>
<td>Bulk concentration of component (j) (mol/m³)</td>
</tr>
<tr>
<td>(c_{ps})</td>
<td>Concentration of species (j) inside the particle pores (mol/m³)</td>
</tr>
<tr>
<td>(D_j)</td>
<td>Effective diffusivity of species (j) inside the catalyst pores (m²/s):</td>
</tr>
<tr>
<td>(D_{A,j})</td>
<td>Diffusivity of component (A) in the mixture (m²/s)</td>
</tr>
<tr>
<td>(E_i)</td>
<td>Activation energy (J/mol)</td>
</tr>
<tr>
<td>(f_{DA})</td>
<td>Deactivation parameter</td>
</tr>
<tr>
<td>(f_{DA,∞})</td>
<td>Deactivation parameter at infinite time</td>
</tr>
<tr>
<td>(k_i)</td>
<td>Reaction rate constant (mol/gm.cat. min)</td>
</tr>
<tr>
<td>(k_{deact})</td>
<td>Catalyst deactivation coefficient</td>
</tr>
<tr>
<td>(K_i)</td>
<td>Adsorption constant</td>
</tr>
<tr>
<td>(N)</td>
<td>Components number</td>
</tr>
<tr>
<td>(M)</td>
<td>Experiment number</td>
</tr>
<tr>
<td>(M_{oj})</td>
<td>Molecular weight of component (j) (g/mol)</td>
</tr>
<tr>
<td>(p)</td>
<td>Pressure (N/m³)</td>
</tr>
<tr>
<td>(P_i)</td>
<td>Partial pressure of (i) component</td>
</tr>
<tr>
<td>(R_g)</td>
<td>Universal gas constant (8.314 J/mol K)</td>
</tr>
<tr>
<td>(R_{r_j})</td>
<td>Net rate of the production of species (j)</td>
</tr>
<tr>
<td>(r)</td>
<td>Particle radial position (m)</td>
</tr>
<tr>
<td>(r_j)</td>
<td>Reaction rate of the production of species (j) (mol/gm.cat min)</td>
</tr>
<tr>
<td>(r_p)</td>
<td>Particle radius (m)</td>
</tr>
<tr>
<td>(V_p)</td>
<td>Total volume of the particles (m³)</td>
</tr>
<tr>
<td>(t)</td>
<td>Time (min)</td>
</tr>
<tr>
<td>(t_{cumu})</td>
<td>Cumulative time for catalyst exposure</td>
</tr>
<tr>
<td>(T)</td>
<td>Temperature (K)</td>
</tr>
<tr>
<td>(V_{Liq})</td>
<td>Total volume of the reactant mixture (m³)</td>
</tr>
<tr>
<td>(Y_i)</td>
<td>Reaction stoichiometry (–)</td>
</tr>
<tr>
<td>(W_c)</td>
<td>Catalyst weight (gm)</td>
</tr>
<tr>
<td>(w)</td>
<td>Catalyst loading (kg/m³)</td>
</tr>
<tr>
<td>(x_j)</td>
<td>Liquid-phase mole fraction of component (j) (–).</td>
</tr>
</tbody>
</table>

**Greek letters**

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
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</thead>
<tbody>
<tr>
<td>(\phi_i)</td>
<td>Association factor for the component (i) (–)</td>
</tr>
<tr>
<td>(\rho)</td>
<td>Gas density (kg/m³)</td>
</tr>
<tr>
<td>(\mu)</td>
<td>Gas viscosity (g/(m s))</td>
</tr>
<tr>
<td>(\Delta)</td>
<td>Difference (–)</td>
</tr>
<tr>
<td>(\varepsilon)</td>
<td>Void factor (–)</td>
</tr>
<tr>
<td>(\varepsilon_p)</td>
<td>Pellet porosity (–)</td>
</tr>
<tr>
<td>(\rho_{Cat})</td>
<td>Catalyst density (kg/m³)</td>
</tr>
<tr>
<td>(\rho)</td>
<td>Mixture viscosity (kg/m.s)</td>
</tr>
</tbody>
</table>

**Subscripts/Superscripts**

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>(b)</td>
<td>Bulk</td>
</tr>
<tr>
<td>CALC</td>
<td>Cinnamyl alcohol</td>
</tr>
<tr>
<td>CALD</td>
<td>Cinnamaldehyde</td>
</tr>
<tr>
<td>cat</td>
<td>Catalyst</td>
</tr>
<tr>
<td>eff</td>
<td>Effective</td>
</tr>
<tr>
<td>H₂</td>
<td>Hydrogen</td>
</tr>
<tr>
<td>HALC</td>
<td>Hydrocinnamyl alcohol</td>
</tr>
<tr>
<td>HALD</td>
<td>Hydrocinnamaldehyde</td>
</tr>
<tr>
<td>(i)</td>
<td>Radial position inside the catalyst particle</td>
</tr>
<tr>
<td>(j)</td>
<td>Number of components</td>
</tr>
<tr>
<td>(l)</td>
<td>Number of experiments</td>
</tr>
<tr>
<td>(\sigma)</td>
<td>Inlet/initial</td>
</tr>
<tr>
<td>(p)</td>
<td>Catalyst particle</td>
</tr>
</tbody>
</table>

**Data Availability**

All relevant data are included within the article.

**Additional Points**

This manuscript provides a logical example of how the basics
of chemical engineering can be combined with computational
optimization techniques to develop a useful kinetic model,
which can be readily used for practical applications on an
industrial scale. Five kinetic models for cinnamaldehyde
hydrogenation were evaluated and compared. Intraparticle
modelling was developed for the liquid-phase hydrogenation
of cinnamaldehyde over a commercial Pt/SiO₂.

**Conflicts of Interest**

The authors declare that they have no conflicts of interest.
References


