

Kinetic Study of Thermal Decomposition of Sugarcane Bagasse Pseudo-Components at Typical Pretreatment Conditions: Towards the Establishment of a Feasible Primary Biorefining

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Research Article

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1 **Kinetic Study of Thermal Decomposition of Sugarcane Bagasse**
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17 **Abstract**

18 A deeper understanding of the kinetics and thermodynamics parameters of sugarcane
19 bagasse (SCB) thermal degradation could define appropriate conditions for primary
20 biorefining in the production of renewable fuels. In this work, the kinetics of thermal
21 degradation of SCB's high polymers are investigated through thermogravimetric data.
22 From this data, model-free and model fitting methods are used to calculate apparent
23 activation energies (E_a) and other related kinetic parameters. DTG curves present three
24 major peaks associated with pseudo-components (PSEs): PSE 1 (hemicelluloses +
25 extractives and lignin), PSE 2 (cellulose + extractives and lignin), and PSE 3 (lignin +
26 extractives and residual holocelluloses). An essential advance is related to the quantitative
27 interpretation of the degradation process by multi-stage modeling governed by diffusion-
28 controlled reactions and order-based models. The Kissinger-Akahira-Sunose method
29 provided E_a ranges of 124 – 154, 147 – 153, and 230 – 530 kJ·mol⁻¹, while the ranges
30 obtained by the Flynn-Wall-Ozawa method were 120 – 152, 144 – 150, and 232 – 545
31 kJ·mol⁻¹, both for PSE 1, PSE 2, and PSE 3, respectively. Data could support the
32 calculation of many critical operating parameters in biorefinery process, such as the
33 pretreatment minimum temperature. SCB biorefining could lead to a degradation of up to
34 10, 0.5, and 11 % for PSE 1, PSE 2, and PSE 3, respectively, at 473.15 K for 200 min.
35 Thermodynamic parameters (ΔH , ΔG , and ΔS) determined SCB endothermic and non-
36 spontaneous thermal degradation.

37

38 **Keywords:** sugarcane bagasse, biorefining, thermal degradation, TGA, model-free
39 method, model fitting method.

40 1. Introduction

41 A difficult task to produce lignocellulosic-based biofuels relies on the compositional and structural
42 barriers of the lignocellulosic materials, which urges solutions in the biomass refining strategies used to
43 enhance the fermentable sugars' yielding [1]. Selective separation of the major lignocellulosic constituents
44 (cellulose, hemicelluloses, and lignin) is mandatory to rupture the cellulose-hemicelluloses-lignin complex
45 through primary biorefining (pretreatment and biomass conditioning) [2]. Such treatments minimize the
46 lignocellulosic biomass recalcitrance, however, it is necessary to evaluate the impact of the operating
47 conditions of these primary biorefining methods on the carbohydrate outcome from biomass pretreatment
48 for further processing and valuation.

49 Some changes might occur in the intrinsic compositional and physical/chemical structure of the
50 cellulose, hemicelluloses, and lignin after the pretreatments. Including, increase in the accessible surface
51 area, reduction in the crystallinity degree of cellulose, removal of hemicelluloses and lignin polymers [3].
52 However, most of the published literature shows insufficient information on the effects of the pretreatment
53 operating conditions on the loss of its major constituents due to thermal degradation. Interestingly, some
54 research groups observed that the concentration of monomers detected does not match the solubilized
55 polysaccharides concentration after pretreatment of sugarcane bagasse [4, 5]. These results revealed a lower
56 yield of fermentable sugar according to certain operating conditions (temperature, residence time, and acid
57 concentration) during the biomass pretreatment [6].

58 Several functions for the most common mechanisms in solid-state reactions such as random
59 nucleation and growth of nuclei, phase boundary controlled reaction, dimensional diffusion, and reaction-
60 order models have been proposed [7, 8]. For decades, researchers overlooked these models regarding
61 studying thermal degradation, mainly adopting the reaction-order models to describe this process [9, 10].
62 However, some analyses forced kinetic data into empirical reaction-order model obtaining an
63 unappropriated representation of the degradation process. The significant variation of apparent activation
64 energies and pre-exponential factors clearly show that thermal degradation involves multiple reactions that
65 should be handled through a model of multiple reactions in parallel.

66 A complicating factor in this regard is that considerable losses of the major biomass constituents
67 occur due to thermal degradation during the primary biorefining [11, 12]. A deeper understanding of the
68 kinetics and thermodynamics related to that type of degradation could support the calculation of the critical
69 operating parameters, such as the pretreatment minimum temperature [13]. In addition, this knowledge
70 would lead to a prospective study to overcome the structural and compositional barriers of lignocellulosic
71 material during biorefining [14]. The experimental and simulation studies could focus on the following
72 aspects:

- 73 • Study of the effects of different pretreatments on the compositional characteristics of the
74 pseudo-components (PSEs), which can be designated as PSE 1 (hemicelluloses + extractives
75 and lignin), PSE 2 (cellulose + extractives and lignin), and PSE 3 (lignin + extractives and
76 residual holocellulose);
- 77 • Reduction of PSEs loss due to thermal degradation, which leads to a decrease in yields of
78 fermentable sugars (from cellulose and hemicelluloses);

- Avoid discrepancies in material balances.

The centerpiece in the kinetic modeling of the thermal degradation is critical to predict the sugarcane bagasse's behavior under various operating parameters during the primary biorefining. It also collaborates for the understanding of the thermal degradation process and clarifies the loss of its major constituents (here referred to as PSEs) [15]. The acceptance of non-isothermal methods is highly motivated because just one non-isothermal experiment can provide the same information as an isothermal method, which usually employs a set of experiments at different temperatures [16]. Besides, the isoconversional models have been used successfully for the estimation of kinetic parameters of several materials [17, 18]. Then, an adequate systematic kinetic analysis of the solid-state mechanism must be resultant of an evaluation of the thermal degradation kinetics under various temperature programming (**Table 1**), enabling detail of the kinetic parameters of this process and also evaluating the results' reliability [19].

This study investigated the kinetics of thermal degradation of the major constituents from sugarcane bagasse (SCB) by three parallel one-step multi reactions in apparent kinetic models. Three model-free non-isothermal methods (Kissinger, Kissinger-Akahira-Sunose (KAS), and Flynn-Wall-Ozawa (FWO) methods) were used to estimate the apparent activation energy ($E_{a,i}$) and Arrhenius pre-exponential factor (A_i) as a function of conversion degree (α). Although the most significant contribution of this work is the estimation of kinetic parameters of the degradation process, there is no doubt that the degradation process does not strictly hold a single-step approximation and also that it might reveal a significant variation in their kinetic parameters. Several solid-state kinetic models (**Table 1**) are analyzed and compared to describe the degradation mechanism and transform that mathematically into a rate equation. The thermodynamic analysis also confirmed the complexity of the SCB thermal degradation mechanism. Finally, carbohydrate and lignin losses due to the thermal degradation process are assessed under certain operating conditions (temperature and residence time) looking for a sustainable primary biorefining. Moreover, these data can be used to develop computational models for the conversion of biomass, which is important to encourage the development of new biorefinery technologies.

2. Materials and Methods

2.1. *Materials and Sample Preparation*

Sugarcane bagasse (SCB) with a moisture content of 52.2 wt % was kindly provided by the São João sugar mill, Araras, São Paulo, Brazil. The moisture content was reduced to 5.0 – 5.5 wt % after five days of natural drying under atmospheric conditions. The SCB was then grounded using an A-11 basic analytical mill equipped with a cutting blade (for pulverizing soft, fibrous grinding materials, IKA® Werke Staufen/Germany), and sieved to a maximum particle size of 0.21 mm. The grounded samples were dried in a drying oven with air renewal circulation (SP LABOR, Brazil) at 378.15 K (± 1.0) for 24 h prior to further analyses. The composition in terms of structural carbohydrates and lignin (41.8 ± 3.3 % cellulose; 31.0 ± 2.5 % hemicelluloses; 17.1 ± 3.4 % lignin; 2.5 ± 0.1 % ash; 9.0 ± 0.1 % extractives) was determined according to the standard National Renewable Energy Laboratory (NREL) protocols based on a two-step acid hydrolysis [20]. The samples were kept in a desiccator at room temperature before the thermogravimetric analysis.

118 2.2. *Thermogravimetric Analysis*

119 The thermogravimetric analysis was carried out under dynamic conditions using a Shimadzu TGA-
120 50 Series (Shimadzu Corporation, International Marketing Division, Tokyo-Japan), equipped with a
121 microbalance and a high-temperature furnace (measurement temperature range: ambient to 1273.5 K). The
122 analyzer was operated from a computer equipped with TA-50 WSI thermal analysis workstation. The mass
123 and temperature calibrations of the TGA-50 Series analyzer were performed according to the
124 manufacturer's recommendations. The milled samples were conditioned in platinum crucibles, each sample
125 weighing 16.00 ± 0.5 mg. The furnace was purged with an inert atmosphere (N_2 flow rate = $50 \text{ mL}\cdot\text{min}^{-1}$)
126 at room temperature and atmospheric pressure. The experiments were carried out from room temperature
127 to 378.15 K using a heating rate of $10 \text{ K}\cdot\text{min}^{-1}$, remaining at the final temperature (378.15 K) for 10 min.
128 This was implemented to ensure that the run began with a dry sample. The isothermal hold at 378.15 K also
129 guaranteed that all samples started with the same temperature distribution and thermal equilibrium.
130 Subsequently, the samples were heated from 378.15 K to 1173.15 K at four programmed heating rates: 10,
131 20, 30, and $40 \text{ K}\cdot\text{min}^{-1}$.

132 2.3. *Signal Processing (Deconvolution)*

133 The derivative thermogravimetric (DTG) curves were deconvoluted into three different peaks
134 (PSE 1, PSE 2, and PSE 3) using iterative fits of the Gaussian distribution function. This procedure allowed
135 obtaining information about the reaction mechanisms and a complete kinetic description through the
136 estimation of the kinetic parameters. Peak deconvolution was performed using the Levenberg-Marquardt
137 algorithm in Microcal Origin® 8.0 software (MicroCal Inc., Northampton, MA, USA), and deviations up to
138 $\pm 8 \%$ were obtained when compared to the experimental data.

139 2.4. *Kinetic, Theoretical Aspects, and Thermodynamics Parameters* 140 *Calculation*

141 The kinetic parameters of the thermal degradation of SCB (based on that of the PSE) were
142 calculated from the TG and DTG curves. Since SCB contains heterogeneous materials (biomass-derived
143 polymers), we believe that the degradation process involves numerous reactions occurring separately, due
144 to the biomass composition complexity. It is thus impossible to know the details of individual reactions.

145 Generally, the thermal degradation can be expressed by the overall reaction described in **Figure**
146 **1**. A parallel independent reaction model could describe the decomposition of SCB. The degradation of the
147 individual SCB pseudo-components can be expressed by:

$$\frac{d\alpha}{dt} = A_i e^{\left(\frac{E_{a,i}}{RT}\right)} f(\alpha) \qquad \text{Equation 1}$$

148 Where α is the degree of conversion, t is the reaction time, A_i and $E_{a,i}$ are the pre-exponential factor,
149 and the activation energy of each reaction in **Figure 1**, respectively, R is the universal gas constant, T is the
150 absolute temperature, and $f(\alpha)$ is the reaction model.

151 The degree of conversion (α) represents the decomposed amount of biomass and it can be defined
152 as:

$$\alpha = \frac{m_0 - m_t}{m_0 - m_f} \quad \text{Equation 2}$$

153 Where m_0 is the sample initial mass, m_t is the sample mass at time t , and m_f is the sample final
154 mass.

155 Considering a constant heating rate (β , commonly used in TG experiments) where T changes
156 linearly with t : $\beta = dT/dt$, Equation 1 is rearranged as follow:

$$\frac{d\alpha}{dT} = \frac{A_i}{\beta} e^{\left(\frac{E_{a,i}}{RT}\right)} f(\alpha) \quad \text{Equation 3}$$

157 Integrating Equation 3 with respect to the temperature yields:

$$g(\alpha) = \int_0^\alpha \frac{d\alpha}{f(\alpha)} = \frac{A_i}{\beta} \int_0^T e^{\left(\frac{E_{a,i}}{RT}\right)} dT \quad \text{Equation 4}$$

158 Different solid-state kinetic models in their integral and differential forms are presented in **Table**
159 **1** [21, 22].

160 2.4.1. Parameter Estimation Procedure: Model Free-Method

161 The model-free method assumes that the thermal degradation reaction rate is only a function of
162 temperature (T). The parameters $E_{a,i}$ and A_i that characterize each reaction, in **Figure 1**, are constant for a
163 specific β and α . Hence, the model free-method allows one to estimate the kinetic parameters $E_{a,i}$ and A_i at
164 a specific α , for an independent kinetic function. Repeating this procedure throughout the conversion degree
165 range, it is possible to obtain a profile of the activation energy as a function of α . Otherwise, the Kissinger
166 method considers the condition of maximum conversion rate to estimate the parameters.

167 **Kissinger Method**

168 The kinetic parameters determined by the Kissinger method are based on the study of the rate
169 equation at the maximum reaction rate, which means that $\frac{d^2 \alpha}{dt^2}$ is equal to zero Equation 5 [23].

$$\frac{d^2 \alpha}{dt^2} = \left(\frac{E_{a,i} \beta}{RT_m^2} + A_i f'(\alpha_m) \exp\left(\frac{-E_{a,i}}{RT_m}\right) \right) \left(\frac{d\alpha}{dt} \Big|_{T_m} \right) = 0 \quad \text{Equation 5}$$

170 Where T_m is the temperature at the maximum peak on the DTG curve, α_m is the conversion degree,
171 and $T_m, \left. \frac{d\alpha}{dt} \right|_{T_m}$ is the maximum reaction.

172 The Kissinger method is also based on the following equation:

$$\ln\left(\frac{\beta}{T_m^2}\right) = \ln\left(\frac{-A_i R}{E_{a,i}} f'(\alpha_m)\right) - \frac{E_{a,i}}{R} \frac{1}{T_m} \quad \text{Equation 6}$$

173 Where the $E_{a,i}$ is determined from the DTG data at different heating rates by linear regression of
174 the $\ln\left(\frac{\beta}{T_m^2}\right)$ versus T_m^{-1} plot. The A_i is estimated using the expressions $f(\alpha)$ and their first derivatives ($f'(\alpha)$),
175 for the kinetic models used to describe the solid-state reactions in **Table 1**.

176 **Kissinger-Akahira-Sunose (KAS) Method**

177 The KAS method for determining the $E_{a,i}$ dependence on α ($E_{a,i\alpha}$) is presented in Equation 7 [24].

$$\ln\left(\frac{\beta_j}{T_{\alpha,j}^2}\right) \cong \left(\ln\frac{RA_{i,\alpha}}{E_{a,i\alpha}} - \ln g(\alpha)\right) - \frac{E_{a,i\alpha}}{RT_{\alpha,j}} \quad \text{Equation 7}$$

178 The plot of $\ln\left(\frac{\beta_j}{T_{\alpha,j}^2}\right)$ versus $\frac{1}{T_{\alpha,j}}$, obtained from experiments recorded at several heating rates (β_i),
 179 should be a straight line with a slope of $-\frac{E_{a,i\alpha}}{R}$.

180 Flynn-Wall-Ozawa (FWO) Method

181 Using Doyle's approximation $\ln p(\alpha) \cong -5.3305 - 1.0516\alpha$ in the α interval from 0.2 - 0.6,
 182 an approximate integral method for determining the dependence of $E_{a,i}$ on α ($E_{a,i\alpha}$) was based on the method
 183 developed by FWO [25].

$$\ln \beta_j \cong -1.0516 \frac{E_{a,i\alpha}}{RT_{\alpha,j}} + \left(\ln \frac{E_{a,i\alpha} A_{\alpha,i}}{R} - \ln g(\alpha) - 5.3305\right) \quad \text{Equation 8}$$

184 Thus, from the plot of the left side of Equation 8 versus $\frac{1}{T_{\alpha,j}}$, obtained from the curves recorded
 185 during multiple constant heating rates (β_j) at constant conversion, the slope can be used to evaluate the
 186 activation energy (approximately $-1.0516 \frac{E_{a,i\alpha}}{RT_{\alpha,j}}$).

187 In both cases, that is the KAS and FWO methods, $g(\alpha)$ is the integral form of the kinetic models
 188 used to describe the solid-state mechanism shown in **Table 1**; $A_{\alpha,i}$ is the pre-exponential factor at constant
 189 α ; $E_{a,i\alpha}$ is the activation energy at constant α .

190 2.4.2. Thermodynamic Parameters

191 Data and information from using the Kissinger, FWO, and KAS model-free methods allowed the
 192 estimation of the thermodynamic parameters of SCB thermal degradation (based on PSE), including the
 193 changes of enthalpy (ΔH), free Gibbs energy (ΔG), and entropy (ΔS) [26].

194 The thermodynamic parameters (ΔH , ΔG , and ΔS), in Equation 9 – 11, were estimated at each β
 195 using the Kissinger method. Otherwise, they were estimated at $10 \text{ K}\cdot\text{min}^{-1}$ to reduce the effect of the
 196 interaction during the thermal degradation [27].

$$\Delta H = E_{a,i} - RT_{\alpha} \quad \text{Equation 9}$$

$$\Delta G = E_{a,i} + RT_m \ln\left(\frac{K_B T_m}{h A_i}\right) \quad \text{Equation 10}$$

$$\Delta S = \frac{\Delta H - \Delta G}{T_m} \quad \text{Equation 11}$$

197 Where, K_B represents the Boltzmann constant ($1.381 \times 10^{-23} \text{ J}\cdot\text{K}^{-1}$), h is the Plank constant
 198 ($6.626 \times 10^{-34} \text{ J}\cdot\text{s}$), and T_{α} is the temperature at specific α .

199 2.4.3. A Kinetic Model for Thermal Degradation of the Major Constituents from 200 SCB

201 The kinetic model corresponding to three parallel one-step multi reactions in **Figure 1** can be
 202 described by the following system of ordinary differential equations:

$$\frac{d \alpha_1}{d t} = A_{\alpha,1} e^{\left(\frac{-E_{\alpha,1}}{RT}\right)} f(\alpha_1) \quad \text{Equation 12}$$

$$\frac{d \alpha_2}{d t} = A_{\alpha,2} e^{\left(\frac{-E_{\alpha,2}}{RT}\right)} f(\alpha_2) \quad \text{Equation 13}$$

$$\frac{d \alpha_3}{d t} = A_{\alpha,3} e^{\left(\frac{-E_{\alpha,3}}{RT}\right)} f(\alpha_3) \quad \text{Equation 14}$$

203 Where $f(\alpha_i)$ is the form of the function describing the reaction model (**Table 1**). The $E_{\alpha,i}$ and $A_{\alpha,i}$
 204 are kinetic parameters estimated from the model-free FWO method as a function of α .

205 The overall degradation rate can be obtained by summing up the rate equations of individual
 206 reactions as Equation 15:

$$\frac{d \alpha}{d t} = \sum_{i=1}^3 \lambda_i A_i \exp\left(\frac{-E_{\alpha,i}}{RT}\right) f(\alpha) \quad \text{Equation 15}$$

207 Where λ_i is the mass loss contribution of constituent i , which is defined as the relative fraction of
 208 the area under the i^{th} DTG PSE peak.

209 The overall DTG curve fitting can be achieved by adding DTG PSE peaks by minimizing the
 210 objective function (OF) through the non-linear least square method defined as Equation 16 [28]:

$$OF = \sum \left(\left. \frac{d \alpha}{d t} \right|_{exp} - \left. \frac{d \alpha}{d t} \right|_{cal} \right)^2 \quad \text{Equation 16}$$

211 Where $\left. \frac{d \alpha}{d t} \right|_{exp}$ and $\left. \frac{d \alpha}{d t} \right|_{cal}$ are experimental and calculated values by the proposed model
 212 conversion rates, respectively.

213 The quality of the model-fit can be determined in terms of *fit (%)* as Equation 17

$$fit(\%) = \left[1 - \frac{\sqrt{\overline{OF}}}{\left(\left. \frac{d \alpha}{d t} \right|_{exp} \right)_m} \right] \times 100 \quad \text{Equation 17}$$

214 Where \overline{OF} is the average of the OF, and $\left(\left. \frac{d \alpha}{d t} \right|_{exp} \right)_m$ is the peak maximum value of the
 215 degradation rate.

216 The iterative calculation process used to solve ordinary differential equations was carried out by a
 217 discretization procedure for the time (as the independent variable) with a minimum value of 101 points,
 218 which gives a trade-off between higher-accuracy discretization. The resulting system was solved using the
 219 fifth/fourth-order Runge Kutta Fehlberg (RKF45) numerical method. The iterative computation process
 220 was implemented using the FORTRAN 90 language, and the FORTRAN code was compiled and executed
 221 by a Compaq[®] FORTRAN compiler to get the outputs of the degraded fraction from the major constituents
 222 from SCB.

223

224 3. Results and discussion

225 3.1. Thermogravimetric Analysis and Deconvolution Peaks

226 The differential mass loss (DTG) curves recorded for SCB at four heating rates (10, 20, 30, and
227 40 K·min⁻¹) under an atmosphere (N₂) are presented in **Figure 2**. As expected, the SCB mass diminished
228 with heating as a result of thermal degradation. Four zones were identified, where several reactions took
229 place. The DTG results show an initial zone where the light volatiles were liberated, and the water
230 evaporation process occurred at temperatures below 450 K. Following, there was thermal degradation of
231 the major constituents from biomass, which proceeded from approximately 450 to 711 K (at 10 K·min⁻¹,
232 **Figure 2a**), 450 to 722 K (at 20·K·min⁻¹, **Figure 2b**), 450 to 733 K (at 30 K·min⁻¹, **Figure 2c**), and 450 to
233 744 K (at 40·K·min⁻¹, **Figure 2d**). Two overlapping peaks, related to two PSEs, could be identified in the
234 main degradation zone. Moreover, above 711 K for the lower heating rate, and above 744 K for the higher
235 heating rate, another zone known as the minor degradation zone, related to PSE 3, could be visualized.

236 The degradation process proceeded faster from 450 to 700 K (second step) with steep slopes of
237 the TG curves were noted. The last remarkable steepness explains the significant loss in mass, due to both
238 faster degradation of the PSE 1 and PSE 2, and the liberation of volatile hydrocarbons. The loss in mass
239 then decreased up to 1173.15 K. The experimental results reported that the solid residue yields were about
240 17.50 % for SCB.

241 In an effort to better identify the zones related to the thermal degradation of the biomass
242 constituents (associated with the PSEs) and their overlapping kinetics at four heating rates, the
243 thermogravimetric data were processed by “deconvolution” of the overlapping peaks by means of the
244 Levenberg–Marquardt’s non-linear optimization method using the gauss multi-peak equation. The
245 comparison of the experimental DTG curves with the results after the deconvolution procedure is shown in
246 **Figure 2**. The adjusted R squared (adjusted-R²) values were above 0.99 for each heating rate (data not
247 shown in this work). SCB is comprised mainly of hemicelluloses, cellulose, and lignin as the major
248 constituents. These biomass polymers were identified at different stages of mass loss as depicted in **Figure**
249 **3**. The peaks show the mass loss rate represented by the DTG curves, recorded during the degradation
250 process of the PSEs at different heating rates. The degradation process described in **Figure 1** consisted of
251 three parallel one-step multi reactions (Rxn) as follows:

- 252 • Rxn1: Associated with hemicelluloses + extractives and lignin. The degradation process
253 occurs at a temperature range from 500 to 700 K with a maximum rate between 605 and 636
254 K, depending on the heating rate (**Figure 3a**). Researchers have reported that hemicelluloses
255 typically decompose in the range of 433 to 613 K [29, 30].
- 256 • Rxn2: Associated with cellulose + extractives and lignin. The degradation process occurs at a
257 temperature range from 650 to 750 K with a maximum rate between 656 and 690 K depending
258 on the heating rate (**Figure 3b**). Furthermore, it is important to highlight that PSE 2 mainly
259 degrades at a higher temperature and narrower temperature zone than PSE 1, suggesting that
260 the degradation rate of PSE 2 is more sensitive to temperature than PSE 1 [31, 32].
- 261 • Rxn3: Associated with lignin + extractives and residual holocellulose (hemicelluloses +
262 cellulose). The degradation process occurs at a slower rate (when compared to the mass losses

263 of PSE 1 and PSE 2) over a much wider temperature range, from 400 to 1050 K, with a
264 maximum rate between 683 and 691 K depending on the heating rate (**Figure 3c**) [31].

265 Thus, the sequence of thermal degradation was: PSE 1 (at 450 – 700 K) > PSE 2 (at 650 – 700) >
266 PSE 3 (at 450 – 1050 K). Nevertheless, the degradation of PSE 3 occurred from 450 up to 1050 K, due to
267 its more thermally stable nature as compared to PSE1 and PSE 2. On the other hand, PSE 2 degraded at
268 higher temperatures but in a narrower temperature zone than PSE 1, suggesting that the degradation rate of
269 PSE 2 was more sensitive to temperature than PSE 1. Finally, the total mass loss at the final temperature of
270 1050 K was 82.50 %.

271 Besides considering individually these PSEs from biomass, they have a biological function in the
272 plant cells, as their organization consists of celluloses microfibers involved with hemicelluloses, altogether
273 surrounded by lignin, forming a complex tridimensional structure that enhances the plant mechanical
274 strength [33]. An explanation of the PSEs' thermal behavior relies on their constituents (PSEs) [34].
275 Hemicelluloses have branched pentoses in their polymeric chain, implying a lower thermal temperature
276 degradation [35]. While cellulose is major constituted of hexoses, in a linear arrangement, resulting in
277 greater thermal stability in comparison to hemicelluloses, however as a result of homogeneous chain
278 structure and high crystallinity, it has a narrower thermal profile [36]. On the other hand, lignin that
279 intensifies the recalcitrance of the biomass, is a heterogeneous, phenolic polymer, with highly aromatic
280 chains, that presents the most thermal stability among the PSEs [37, 38]. Another assumption that can be
281 made for these PSEs is that due to the crystalline structure of cellulose, this PSE 1 directly affects the kinetic
282 parameters ($E_{a,i}$) since more heat is required to thermal activate this highly organized component.
283 Furthermore, the amorphous structure of lignin confers the final tail at the TG curve profile, lowering its
284 PSE 3 - $E_{a,i}$ to a wide temperature range, to form the lignin char [39].

285 The kinetic analysis of the thermal degradation was carried out by model-free methods as
286 described earlier, based on the fact that an understanding of the degradation process from the biomass PSEs
287 is required to determine the loss of major constituents from the SCB under the operational conditions of
288 time and temperature.

289 **3.2. Estimation of Kinetic Parameters and Mechanism Analysis**

290 The properties of three peaks obtained at four heating rates (10, 20, 30, and 40 K·min⁻¹) were
291 summarized in **Table 2**. The results demonstrated that all the peaks or deflections shifted proportionally
292 towards higher reaction temperatures with the heating rates, which means that the peak temperature (T_m),
293 peak height (H_m), and conversion at T_m (α_m) were dependent on the heating rate. Thus, the variations in the
294 properties of the three peaks with the heating rate account for the three parallel one-step multi reactions of
295 PSE 1, PSE 2, and PSE 3, respectively (**Figure 1**).

296 The model-free method proposed by Kissinger to calculate the $E_{a,i}$, and A_i , from the DTG data at
297 four heating rates (10, 20, 30, and 40·K min⁻¹) was used in this work. The kinetic data were obtained from
298 the peaks of the DTG curves (**Figure 3**), assuming different kinetic functions ($f(\alpha)$) to estimate A_i .

299 According to this assumption, the plot $\ln\left(\frac{\beta}{T_m^2}\right)$ versus T_m^{-1} should be a straight line, where the slope
300 allows one to calculate the $E_{a,i}$. This plot was constructed for each PSE and the results are shown in **Figure**

301 **4**.

302 The data obtained for the kinetic parameters ($E_{a,i}$ and A_i) were statistically analyzed to find the
303 degree of reliability, and the results can be seen in **Table 2**. The values obtained for the $E_{a,i}$ and A_i of the
304 three PSEs were derived from the slope and intercept, respectively, and were all of linear fit. Regarding the
305 estimation of the $E_{a,i}$, it is clear that this is a positive value and does not depend on the heating rate. The $E_{a,i}$
306 values were within the ranges reported for each PSE in **Table 2**. For the PSE 1, PSE 2, and PSE 3, the
307 values obtained by the Kissinger method for the $E_{a,i}$ were about 135; 139; and 328 kJ·mol⁻¹, respectively.
308 Hence the sequence for the $E_{a,i}$ of the biomass PSEs is: PSE 3 > PSE 2 > PSE 1. The A_i estimated for the
309 thermal degradation of the SCB's PSEs varied over a wide range, since six kinetic functions, $f(\alpha)$, were
310 used (R2, R3, F1, D2, D3, and A2, in **Table 1**) as shown in **Table 2**.

311 To confirm the values obtained using the Kissinger method and avoid erroneous values in the set
312 of parameters obtained for the $E_{a,i}$ and A_i using the different kinetic functions, $f(\alpha)$, the kinetic results were
313 compared with the data obtained using the FWO and KAS methods.

314 Thus, the model-free method established a dependence of the $E_{a,i}$ on the conversion rate, α [40].

315 The quantitative kinetic analysis carried out using iso-conversional methods is based on
316 conversion–temperature data, thus the conversion degree (α) plot against the temperature produces sigmoid
317 curves **Figure 3**.

318 In the α versus temperature curves, three different profiles (sigmoid curves) were identified,
319 associated with the kinetics of thermal degradation of the three PSEs. Consequently, three regions
320 confirmed the zones shown in the DT/DTG curves, characterized as lower conversion rates, being no bigger
321 than 2.0, 4.0, and 8.0 % for PSE 1, PSE 2, and PSE 3, respectively. In the next section of the α versus
322 temperature curves, the conversion rate in the ranges 500 – 700 K (for PSE 1), 650 – 750 K (for PSE 2),
323 and 500 – 900 K (for PSE 3), was more intense in the evaluated range, as regards the volatilization of
324 complex organic molecules and/or hydrocarbons. The third stage of these curves appeared at 700 K for PSE
325 1 and PSE 2, and at 900 K for PSE 3, since the conversion with temperature is at its lowest value, probably
326 the degradation process is almost finished.

327 In order to study the thermal effect on the three parallel thermal degradation reactions of the three
328 PSEs at four heating rates (10, 20, 30, and 40 K·min⁻¹), the KAS and FWO model-free methods were used
329 to estimate the kinetic parameters ($E_{a,i\alpha}$ and $A_{\alpha,i}$) as a function of the conversion degree, α .

330 In this study, ten conversion degree values of 0.05 – 0.9 (equivalent to 5 – 90 %) with an increment
331 of 0.1 (or 10 %) after 0.1, under each heating rate condition were analyzed. **Figure 5** shows the estimates
332 for the kinetic parameters, for the whole set of α values used in the linear plot, as determined using the
333 FWO and KAS methods. **Table 3** gives an overview of the adjusted-R² values corresponding to the linear
334 fit data in **Figure 5**, together with the values obtained for the $E_{a,i\alpha}$ and $A_{\alpha,i}$, at each conversion degree
335 (between 0.05 < α < 0.9) in the thermal degradation of PSE 1, PSE 2, and PSE 3.

336 In both cases, it is important to remember that the FWO and KAS methods apply an integral
337 approximation, valid between 0.2 < α < 0.6 and 0.2 < α < 0.5, respectively. Thus, the kinetic parameters
338 estimated out of the valid ranges may not show a physical and/or mathematical sense (being a specific case
339 of kinetic data obtained for the thermal degradation of PSE 3 at a conversion degree greater than 0.6).

340 From the data obtained for the adjusted-R² shown in **Table 3**, it can be seen that both the FWO
341 and KAS methods performed well during the fit of the linear regression data, and the $E_{a,i\alpha}$ values estimated

342 from these two methods are close to each other at the same α . This confirms that the methods studied in
343 this work can be used to calculate the kinetic parameters. This was also demonstrated in studies of the
344 thermal stages of SCB [33], petroleum derived [41], polymers [42], and other biomass types [43].

345 Analyzing Rxn1, related to the thermal degradation of PSE 1, it can be seen that the linear profiles
346 (for the FWO and KAS plots) were nearly parallel in the conversion range of $0.2 < \alpha < 0.8$, which implies
347 the chance of a single degradation reaction mechanism or a multi-reaction mechanism. However, in the
348 ranges below or equal to 0.2 and above or equal to 0.8, the degradation reactions may be different, due to
349 the lack of parallelism of the lines. In consequence, the iso-conversion analyses indicated variations in both
350 $E_{a,i\alpha}$ and $A_{\alpha,i}$ values, with an increase in the conversion degree, α . This observation provides evidence
351 that the degradation process of PSE 1 takes place in multi-steps. In contrast, the degradation process of PSE
352 2 corresponds to a single step reaction or a multi-reaction mechanism, since there was little variation in the
353 $E_{a,i\alpha}$ values (147 – 153 kJ·mol⁻¹ by the KAS method, and 144 – 150 kJ·mol⁻¹ by the FWO method) with an
354 increase in the conversion degree.

355 On the other hand, the thermal degradation rate of PSE 3 was identified as a complex reaction,
356 showing changes in the apparent activation energy with increases in the conversion degree. This change
357 will depend on: (i) the contribution of each compound present in PSE 3 (lignin + extractives and residual
358 holocellulose), (ii) the complexity of the degradation reaction(s), and (iii) the multi-step reactions.

359 **3.3. Estimation of Thermodynamic Parameters**

360 Thermodynamic parameters are significant to adequately design reactors on a larger scale for
361 primary biorefining processes. According to the results of the last section, the $E_{a,i}$ derived from model-free
362 methods (**Table 2** and **Table 3**) were used to calculate the thermodynamic parameters (ΔH , ΔG , and ΔS)
363 (**Table 2** and **Table 3**).

364 The enthalpy (ΔH) changes revealed the energy difference between the reagent and the activated
365 complex [26]. Moreover, a positive ΔH indicates that energy is required for the reactants to reach their
366 transition state, implying that the thermal degradation reactions were all endothermic [26]. As seen in **Table**
367 **2** and **Table 3**, the value of ΔH increases with α . The average values of ΔH for PSE 1, PSE 2, and PSE 3
368 are 141.7, 144.7, and 300.8 kJ·mol⁻¹, respectively, for FWO method; 139.1, 141.3, and 306.4 kJ·mol⁻¹,
369 respectively, for KAS method. The results demonstrated that more heat energies are required for PSE 3
370 degradation process than PSE 1 and PSE 2 to dissociate the reagents bonds, which agrees with $E_{a,i}$ derived
371 from the model-free method.

372 The changes of the Gibbs free energy (ΔG) reflect the energy variation of the reactional system to
373 approximate the molecules and form the activated complex [44], which means that the process is endergonic
374 and not spontaneous in the forward direction. During the thermal degradation, by ΔG it is possible to
375 evaluate the disorder change, where low ΔG values favors the reaction. Furthermore, ΔG has no noticeable
376 change for PSE 1 and PSE 2. The average values of ΔG for PSE 1, PSE 2, and PSE 3 are 323.5, 343.1, and
377 454.5 kJ·mol⁻¹, respectively, for FWO method; 321.1, 339.9, and 457.4 kJ·mol⁻¹, respectively, for KAS
378 method. With these, the favorability order of the degradation process was PSE 1 > PSE 2 > PSE 3. Then,
379 PSE 1 and PSE 2 consumed a great portion of heating in the degradation process, disordering the system,
380 and favoring the degradation process.

381 The entropies (ΔS) had negative variation values (**Table 2** and **Table 3**), indicating a lower degree
382 of disorder of the products compared to the initial PSEs in the thermal degradation process (**Figure 1**).

383 All in all, as thermal degradation is endothermic, $\Delta H > 0$, and the entropy of the system decreases,
384 $\Delta S < 0$, the sign of ΔG is positive at all temperatures. Thus, the degradation process is never spontaneous.

385 **3.4. Thermal Degradation Model with Focus on Primary Biorefining Processes**

386 Based on FWO and KAS method, it is possible to highlight that when $\alpha > 0.05$, it is an indicator
387 that the thermal degradation begins to occur in PSE 1 and the amorphous portion of the PSE 2 structure,
388 allowing an increase in the E_a values. Hence, PSE 1 had a specific reaction heating, where during its
389 degradation, high levels of CO_2 are formed, whereas in lower contents are formed CO, methanol, methane,
390 furfural, 5-hydroxymethylfurfural, and anhydrous sugars produced in substantial amounts from other
391 polysaccharides [45, 46]. The degradation of cellulose, here referred to as PSE 2, is decomposed into
392 levoglucosan, hydroxyacetaldehyde, hydroxylactone, pyruvic aldehyde, glyceraldehyde, 5-hydroxymethyl-
393 furfural, and furfural, involving activation energy lower than $211 \cdot \text{kJ} \cdot \text{mol}^{-1}$ [12, 46]. This indicates that PSE
394 1 and PSE 2 have lower thermal stability, though, PSE 3 has the highest thermal stability, degrading in a
395 wider temperature range.

396 As seen in **Table 1**, the solid-state kinetic models are identified into four groups: A_n , R_n , D_n , and
397 F_n , respectively, these mechanisms describe: random nucleation, nuclei growth, phase boundary controlled
398 reaction, dimensional diffusion, and reaction-order models. Using the average E_a obtained from the model-
399 free method, the mass loss contribution (λ_i), in Equation 15, was estimated and mentioned in **Table 4**. The
400 difference between the experimental and predicted values ($\left. \frac{d\alpha}{dt} \right|_{exp}$ and $\left. \frac{d\alpha}{dt} \right|_{cal}$) are reported as indicated
401 by OF and Fit (%). The simulated $\left. \frac{d\alpha}{dt} \right|_{cal}$ curves as a function of temperature from different solid-state
402 kinetic models are illustrated in **Figure 6** and the corresponding optimized parameters are included in **Table**
403 **4**.

404 Comparing $\left. \frac{d\alpha}{dt} \right|_{exp}$ and $\left. \frac{d\alpha}{dt} \right|_{cal}$, the experimental data overlapped the D2, D3, and F1
405 heterogeneous kinetic models. According to literature, these heterogeneous mechanisms refer to a diffusion
406 process in two and three dimensions, D2 and D3 respectively. They are related to the heat transfer capacity
407 along with the material structure and order-based models representing simple models as they are similar to
408 those used in homogeneous kinetics [47]. Similar results were described by for other biomasses [48]. In
409 addition, thermal degradation follows a reaction-order model (F1) which is the simplest model as it is
410 similar to those used in homogeneous kinetics. However, a misconception may be the result of
411 approximating the degradation of PSEs from SCB towards A2, R2, or R3 heterogeneous kinetic models.

412 In order to advance and confirm the thermal degradation of major constituents associated with PSE
413 1, PSE 2, and PSE3, respectively, during the primary biorefining of SCB, different operating conditions
414 were evaluated (**Figure 7**). Changes in biomass could be observed in the evolution of the profiles of the
415 PSEs and the percent of degradation is used as an indicator of the liberation of thermally degraded fractions
416 (TDF). The values for TDF were between 0.0 - 35 % (**Figure 7a**), 0.0 - 0.6 % (**Figure 7b**) and 0.9 - 16 %
417 (**Figure 7c**) for PSE 1, PSE 2, and PSE 3, respectively, when considering a temperature range from 393.15
418 - 503.15 K, and 240 min.

419 It can be seen in **Figure 7** that each PSE exhibited a very different thermal degradation behavior.
420 PSE 1 (**Figure 7a**), and PSE 3 (**Figure 7c**) tended to produce relatively high levels of TDF up to 100 min.
421 On the other hand, PSE 1 produced significantly less TDF than PSE 3. Nevertheless, at temperatures above
422 483.15 K, the degradation process exhibited relatively high levels of TDF when compared to PSE 1. PSE
423 2 tended to produce relatively low levels of TDF (**Figure 7b**).

424 It is important to mention that higher temperatures than 503 K were not analyzed since this would
425 represent roasting [49], and pyrolysis [40] of the biomass. Thus, a great deal of reaction mechanism
426 information is required to properly model these processes. More accurate and detailed models can be
427 assumed as presented in the published literature [50].

428 Thermal degradation was shown to have a great influence on the major biomass constituents as
429 hemicelluloses (PSE 1), cellulose (PSE 2), and lignin (PSE 3). One of the motivations of this work was to
430 explore how the temperature and time, during primary biorefining, affected the degradation of the biomass
431 constituents. The results showed that the effect of temperature and time on the degradation process was
432 important and should be considered when primary biorefining SCB (i. e. acid-catalyzed hydrothermal
433 pretreatment, hydrothermal pretreatment, organosolv, and ammonia fiber expansion (AFEX)), as they can
434 directly affect the availability of fermentable sugars.

435 In the conception of SCB, a byproduct generated by the agricultural sugar-alcohol industry in the
436 production of biofuels, and an economical alternative for the sustainable production of electricity (co-
437 generation), it is important to minimize TDF, since the thermal degradation of biomass fuels is related to
438 their major constituents (i.e., considering the degradation behavior of cellulose, hemicelluloses, and lignin
439 under the combination of time and temperature in the operating conditions). We believe that TDF below 5
440 % in the primary biorefining process is acceptable in the processing of SCB. In the field of research into
441 primary biorefining, this approach could be useful, since the degradation of the major constituents from
442 biomass is still unknown. This thermal degradation model will be incorporated into the global
443 physicochemical conversion of SCB and material balances.

444

445 **Conclusions**

446 The DTG curves showed that thermal degradation of sugarcane bagasse (SCB) did not occur in a
447 single reaction. This behavior was evidenced by the three parallel one-step multi reaction overlapping peaks
448 corresponding to three pseudocomponents (PSEs): PSE 1 (hemicelluloses + extractives and lignin), PSE 2
449 (cellulose + extractives and lignin), and PSE 3 (lignin + extractives and residual holocellulose). The ranges
450 of the apparent activation energies were specified for each model-free method. The Kissinger method
451 reported approximately 135, 139, and 328 kJ·mol⁻¹, respectively, for PSE 1, PSE 2, and PSE 3; but offered
452 no information about the multi-step or complex reactions taking place during the thermal degradation
453 process. It was, therefore, useful to apply the Flynn-Wall-Ozawa (FWO) and Kissinger-Akahira-Sunose
454 (KAS) model-free methods. The KAS method reported the variability of the apparent activation energy
455 values as 124 – 154, 146 – 153, and 230 – 530 kJ·mol⁻¹; and the FWO method reported values of 120 – 152,
456 144 – 150, and 232 – 545 kJ·mol⁻¹, respectively, for PSE 1, PSE 2, and PSE 3. Thus, the degradation of
457 PSE 3 occurred from as low as 450 K and up to 1050 K, due to its more thermally stable nature as compared

458 to PSE1 and PSE 2. On the other hand, the degradation of PSE 2 was more sensitive to temperature than
459 that of PSE 1. The thermodynamic parameters (positive ΔH , positive ΔG , and negative ΔS) were determined
460 by the activated complex theory. These validated that the degradation process of SCB major constituents
461 was endothermic and non-spontaneous. Consequently, thermal degradation of the major constituents of
462 SCB is governed by diffusion models (i.e., two and three-dimensional diffusion - Jander equation) and
463 reaction-order models (i.e., first-order or unimolecular decay law). The thermally degraded fractions (TDF)
464 in the establishment of sustainable primary biorefining can attain up to 5%. Combining knowledge of the
465 TDF and the material balances it will be possible to minimize the mismatches in the amount of solubilized
466 polysaccharides in front of detected sugar concentrations and recovery of the high polymers (cellulose,
467 hemicelluloses, and lignin) from biomass.

468

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614 **Declaration of Interest**

615 The authors declare that are no financial competing interests.

616 **Authors Contribution**

617 Bahú, J.O.: Conceptualization, Visualization, Writing - Original Draft, Review & Editing;
618 Gariboti, J.C.J., Rivera, E.C., Lopes, E.S., Lopes, M.S., Felisbino, R.F.: Writing - Original Draft; Maciel
619 Filho, R.: Supervision, Project Administration; Tovar, L.P.: Data Collection, Analysis, Conceptualization,
620 Supervision, Writing - Original Draft, Review & Editing. All authors read and approved the final
621 manuscript.

622 **Data Availability**

623 The datasets generated during and/or analyzed during the current study are available from the
624 corresponding author, prof. dr. Laura Plazas Tovar, laura.tovar@unifesp.br, on reasonable request.

Table 1 Models ($f(\alpha)$) used in the heterogeneous kinetics, their first derivatives ($f'(\alpha)$), and the integral form ($g(\alpha) = \int_0^\alpha \frac{d\alpha}{f(\alpha)}$) examined in this work

Reaction model	Code	$f(\alpha)$	$f'(\alpha)$	$g(\alpha)$
<i>Random nucleation and nuclei growth</i>				
Avrami-Erofe'ev	A2	$2 \times (1-\alpha) \times [-\ln(1-\alpha)]^{1-1/2}$	$[2 \times \ln(1-\alpha) + 2 - 1] / \{-\ln(1-\alpha)\}^{1/2}$	$[-\ln(1-\alpha)]^{1/2}$
<i>Phase boundary-controlled reaction</i>				
Contracting area (<i>i.e.</i> , bidimensional shape)	R2	$(1-\alpha)^{1/2}$	$-1/[2 \times (1-\alpha)^{1/2}]$	$[1 - (1-\alpha)^{1/2}]$
Contracting volume (<i>i.e.</i> , tridimensional shape)	R3	$(1-\alpha)^{2/3}$	$-2/[3 \times (1-\alpha)^{1/3}]$	$[1 - (1-\alpha)^{1/3}]$
<i>Diffusion models</i>				
Two-dimensional diffusion	D2	$[-\ln(1-\alpha)]^{-1}$	$-1/\{[\ln(1-\alpha)]^2 \times (1-\alpha)\}$	$[(1-\alpha) \times \ln(1-\alpha)] + \alpha$
Three-dimensional diffusion (Jander equation)	D3	$[3 \times (1-\alpha)^{2/3}] / \{2 \times [1 - (1-\alpha)^{1/3}]\}$	$[0.5 - (1-\alpha)^{-1/3}] / [1 - (1-\alpha)^{1/3}]^2$	$[1 - (1-\alpha)^{1/3}]^2$
<i>Reaction-order models</i>				
First-order (unimolecular decay law)	F1	$(1-\alpha)$	-1	$-\ln(1-\alpha)$

627
628**Table 2** The properties of three peaks of the SCB sample' PSEs obtained at four heating rates (10, 20 30, and 40 K·min⁻¹), and values estimated (apparent activation energy, E_a , and pre-exponential factor, A) using the Kissinger method for the adjusted-R²

Parameters type		PSE1				PSE 2				PSE 3					
Heating rate, β (K·min ⁻¹)		10	20	30	40	10	20	30	40	10	20	30	40		
Peak temperature, T_m (K)		605.8	620.3	631.0	635.7	656.3	672.6	684.2	690.6	683.1	683.9	689.3	691.6		
Peak height, H_m (K ⁻¹)		0.014	0.026	0.038	0.055	0.023	0.039	0.054	0.074	0.001	0.002	0.004	0.006		
Conversion at T_m , α_m (-)		0.488	0.501	0.509	0.514	0.496	0.514	0.523	0.533	0.476	0.487	0.49	0.491		
Kinetic	E_{ai} (kJ·mol ⁻¹)	134.6				138.9				328.1					
		R2	26.3	26.2	26.2	26.2	24.9	24.8	24.8	24.8	58.6	58.6	58.6	58.6	
	ln A_i (min ⁻¹) using different $f(\alpha)$ kinetic functions	R3	26.1	26.1	26.1	26.1	24.7	24.7	24.7	24.7	58.4	58.4	58.4	58.4	
		F1	25.9				24.5				58.2				
		D2	24.4	24.5	24.5	24.5	23.1	23.1	23.2	23.2	56.7	56.7	56.8	56.8	
		D3	23.0	23.0	23.1	23.1	21.6	21.7	21.7	21.8	55.2	55.3	55.3	55.3	
		A2	26.8	26.7	26.6	26.6	25.3	25.2	25.1	25.0	59.2	59.1	59.1	59.1	
		Adjusted-R ²		0.993				0.998				0.963			
		Thermodynamic	ΔH (kJ·mol ⁻¹)	129.5				133.4				322.4			
				R2	311.3	315.7	318.9	320.3	331.1	336.0	339.5	341.4	523.5	523.7	525.3
ΔG (kJ·mol ⁻¹) using different $f(\alpha)$ models	R3		311.3	315.7	318.9	320.3	331.1	336.0	339.5	341.5	523.5	523.8	525.4	526.1	
	F1		311.4				331.2				523.5				
	D2		311.7	316.0	319.2	320.6	331.5	336.4	339.9	341.8	523.7	523.9	525.5	526.2	
	D3		312.0	316.3	319.5	320.9	331.9	336.8	340.3	342.2	523.8	524.1	525.7	526.4	
	A2		311.2	315.6	318.8	320.2	331.0	335.9	339.4	341.4	523.4	523.7	525.3	526.0	
	ΔS (kJ·mol ⁻¹ ·K ⁻¹) using different $f(\alpha)$ models		R2	-300.1	-300.3	-300.4	-300.5	-301.2	-301.4	-301.5	-301.6	-294.4	-294.4	-294.5	-294.5
			R3	-300.1	-300.3	-300.5	-300.5	-301.2	-301.4	-301.6	-301.7	-294.4	-294.4	-294.5	-294.5
			F1	-300.2				-301.3				-294.4			
D2		-300.7	-300.8	-301.0	-301.0	-301.8	-302.0	-302.1	-302.2	-294.7	-294.7	-294.7	-294.8		
D3		-301.2	-301.3	-301.5	-301.5	-302.3	-302.5	-302.6	-302.7	-294.9	-294.9	-294.9	-295.0		
A2		-299.9	-300.1	-300.3	-300.4	-301.0	-301.3	-301.5	-301.5	-294.3	-294.3	-294.4	-294.4		

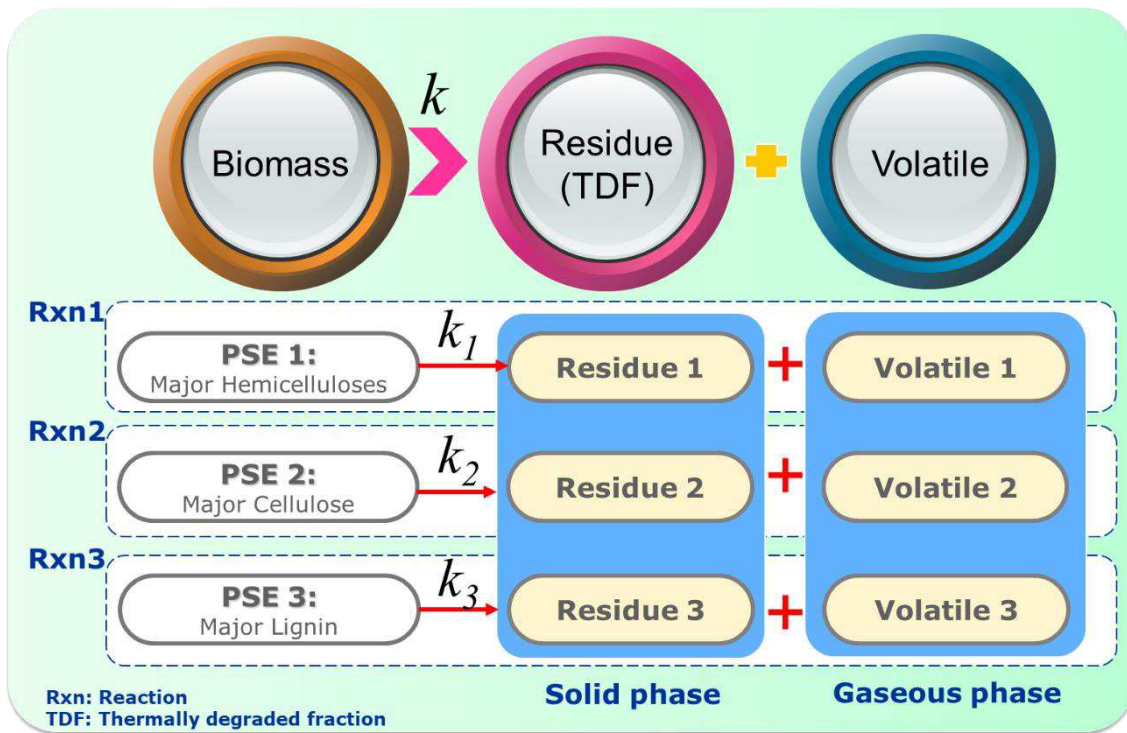
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 A_i in min⁻¹

Table 4 Relative mass loss contributions (λ_i) of SCB constituents at different heating rate (β)

Heterogeneous kinetics model	β (K min ⁻¹)	PSE 1	PSE 2	PSE 3	OF (s ⁻²)	fit (%)
A2	10	0.067	0.265	0.668	1.6×10 ⁻⁶	85.8
	20	0.200	0.791	0.009	4.7×10 ⁻⁶	86.9
	30	0.201	0.797	0.002	9.4×10 ⁻⁶	87.1
	40	0.158	0.627	0.214	5.6×10 ⁻⁵	74.7
	Average (±SD)	0.157 (±0.045)	0.620 (±0.177)	0.223 (±0.222)		
R2	10	0.126	0.590	0.284	1.5×10 ⁻⁶	89.3
	20	0.157	0.603	0.240	5.2×10 ⁻⁶	89.3
	30	0.127	0.590	0.283	9.9×10 ⁻⁶	89.6
	40	0.127	0.591	0.283	1.5×10 ⁻⁵	89.7
	Average (±SD)	0.134 (±0.012)	0.594 (±0.005)	0.272 (±0.016)		
R3	10	0.481	0.258	0.261	1.4×10 ⁻⁶	89.8
	20	0.442	0.315	0.243	2.2×10 ⁻⁶	93.0
	30	0.436	0.326	0.238	3.4×10 ⁻⁶	93.9
	40	0.232	0.389	0.379	5.3×10 ⁻⁶	93.9
	Average (±SD)	0.398 (±0.083)	0.322 (±0.035)	0.280 (±0.050)		
D2	10	0.508	0.305	0.187	7.1×10 ⁻⁸	97.7
	20	0.441	0.320	0.239	1.2×10 ⁻⁶	94.8
	30	0.690	0.201	0.110	6.0×10 ⁻⁶	91.9
	40	0.184	0.287	0.529	9.1×10 ⁻⁶	92.0
	Average (±SD)	0.456 (±0.143)	0.278 (±0.039)	0.266 (±0.131)		
D3	10	0.501	0.333	0.165	7.5×10 ⁻⁸	97.6
	20	0.448	0.325	0.227	1.6×10 ⁻⁶	94.0
	30	0.259	0.719	0.022	4.6×10 ⁻⁶	92.9
	40	0.282	0.591	0.127	7.6×10 ⁻⁶	92.7
	Average (±SD)	0.373 (±0.102)	0.492 (±0.163)	0.135 (±0.061)		
F1	10	0.514	0.374	0.112	4.5×10 ⁻⁷	94.1
	20	0.473	0.339	0.188	1.9×10 ⁻⁶	93.6
	30	0.481	0.354	0.166	4.3×10 ⁻⁶	93.2
	40	0.476	0.373	0.151	8.0×10 ⁻⁶	92.5
	Average (±SD)	0.486 (±0.014)	0.360 (±0.014)	0.154 (±0.023)		

634 **Figure 1** Reaction scheme describing thermal degradation of major constituents from biomass

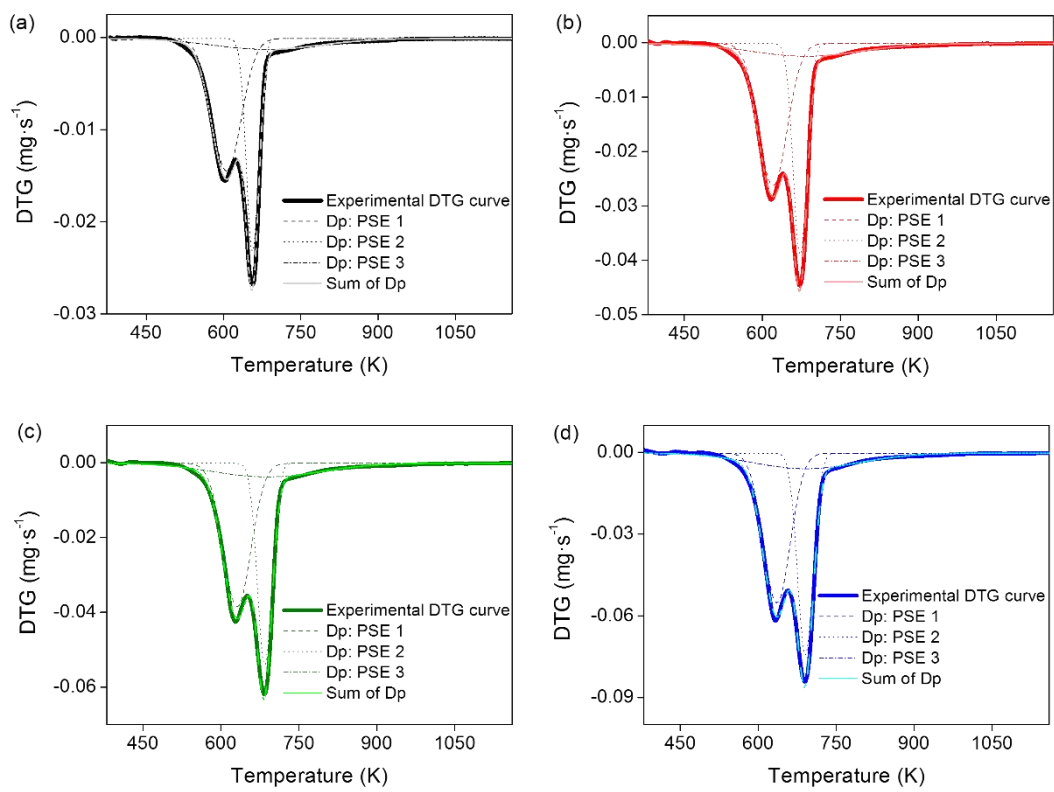


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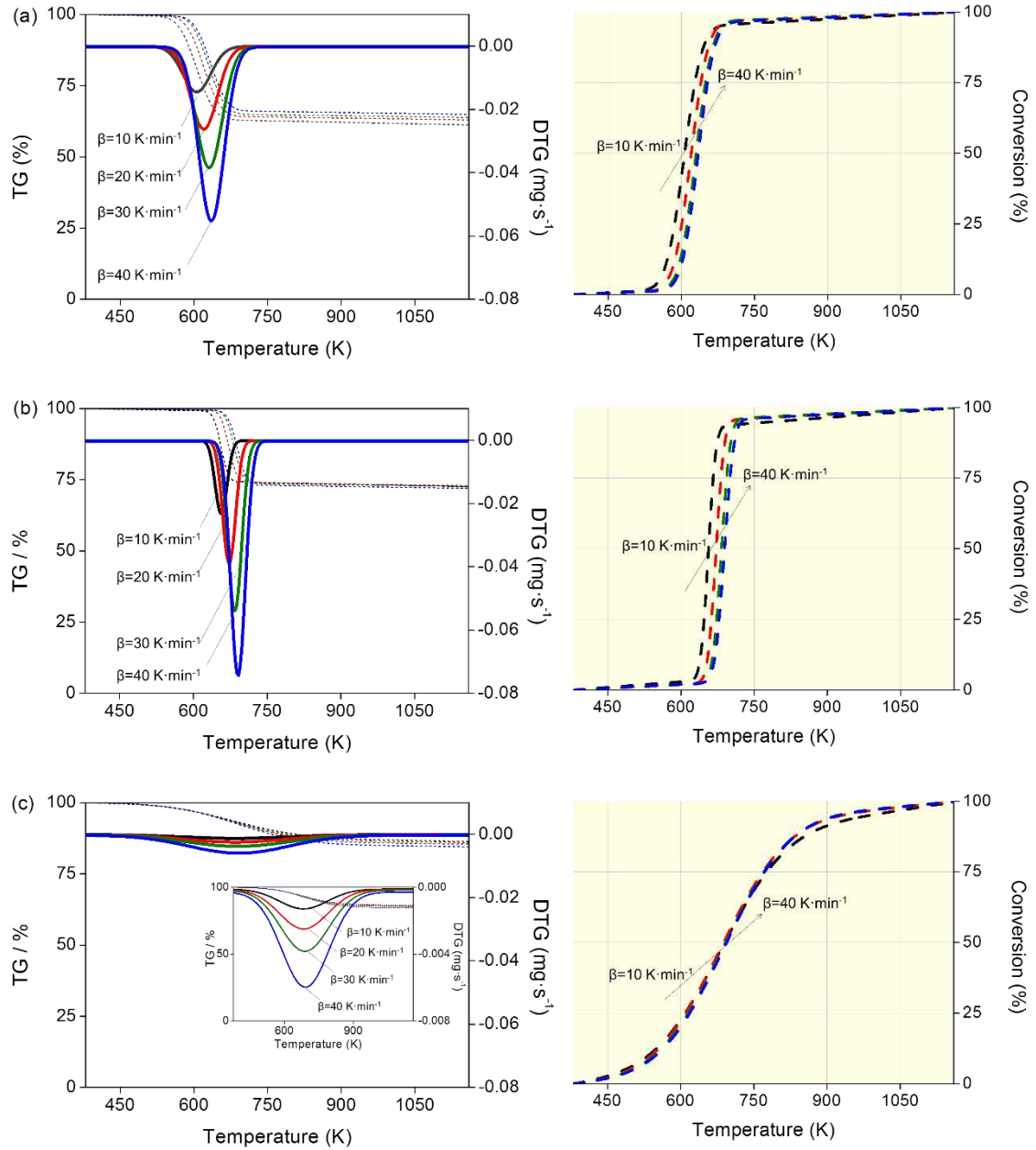
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Figure 2 The deconvolution of three unresolved peaks (Dp) from the DTG curves of SCB samples at four heating rates under an inert atmosphere (N_2): (a) $10\text{ K}\cdot\text{min}^{-1}$, (b) $20\text{ K}\cdot\text{min}^{-1}$, (c) $30\text{ K}\cdot\text{min}^{-1}$, and (d) $40\text{ K}\cdot\text{min}^{-1}$



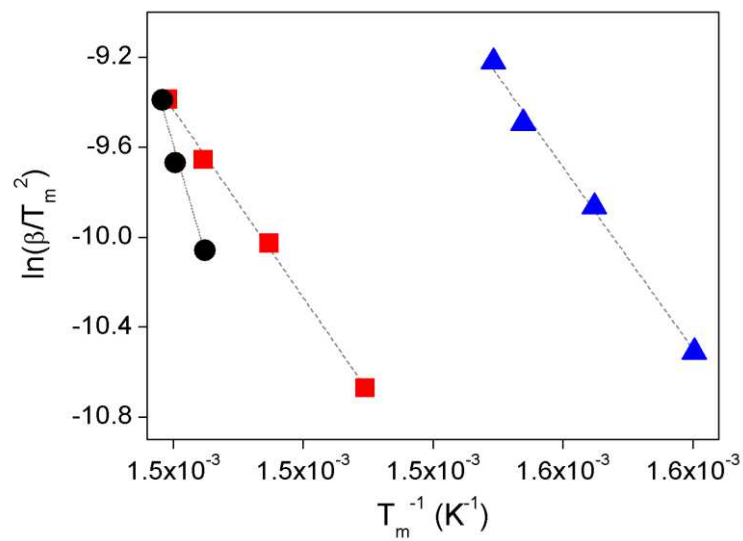
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642 **Figure 3** TG curves (dot lines), DTG curves (continuous lines), and conversion (dash lines) of the high
 643 polymers from SCB recorded at four heating rates: 10 K·min⁻¹ (····, —, ---), 20 K·min⁻¹ (····, —, ---), 30
 644 K·min⁻¹ (····, —, ---) and 40 K·min⁻¹ (····, —, ---) under an inert atmosphere (N₂) for (a) PSE 1:
 645 hemicelluloses + extractives and lignin, (b) PSE 2: cellulose + extractives and lignin, and (c) PSE 3: lignin
 646 + extractives and residual holocellulose (hemicelluloses + cellulose)



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649 **Figure 4** Kissinger plot of the values for $\ln\left(\frac{\beta}{T_m^2}\right)$ as a function of T_m^{-1} . \blacktriangle PSE 1: hemicelluloses+
 650 extractives and lignin; \blacksquare PSE 2: cellulose+ extractives and lignin; \bullet PSE 3: lignin+ extractives and residual
 651 holocellulose (hemicelluloses + cellulose); ---- linear fit

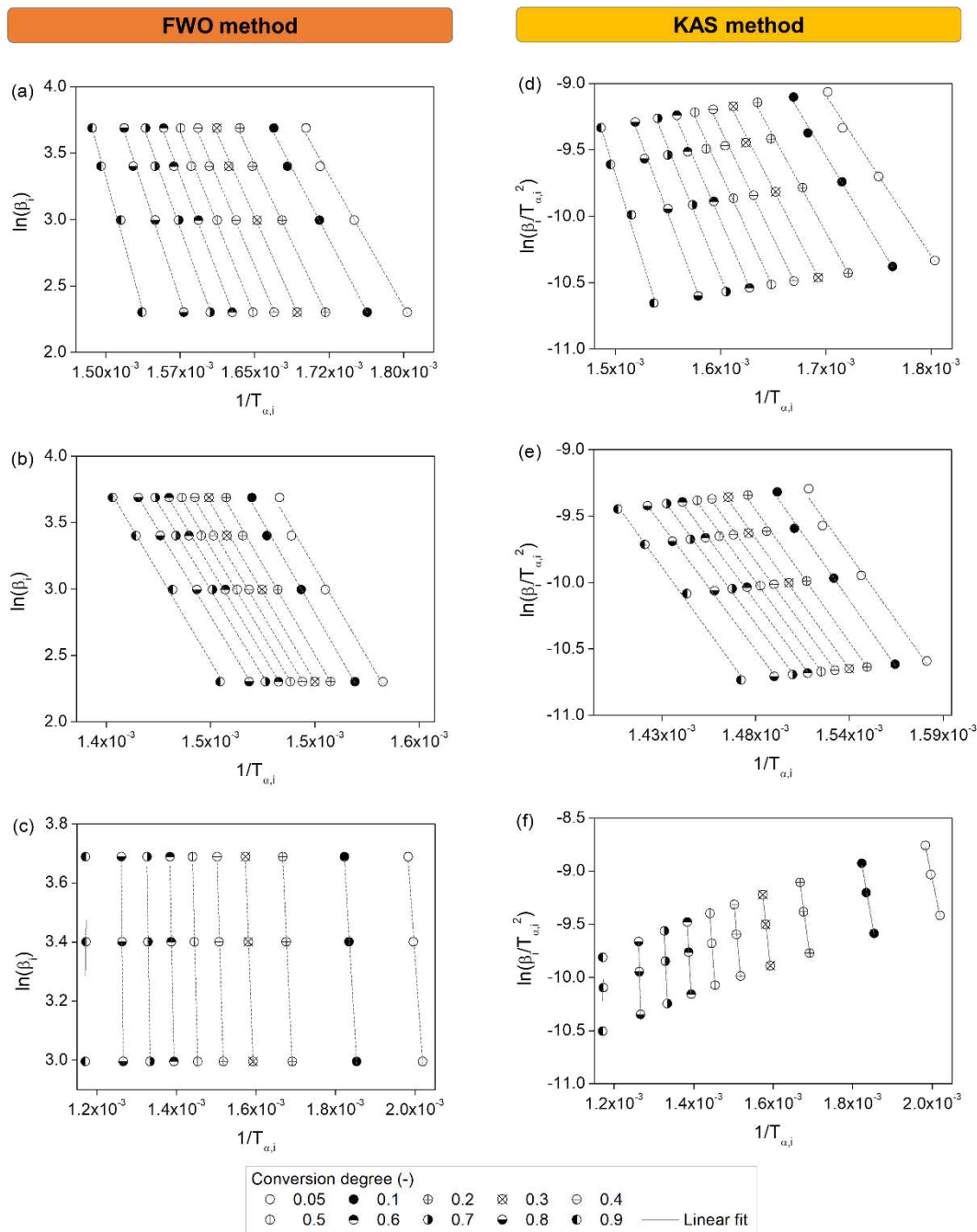


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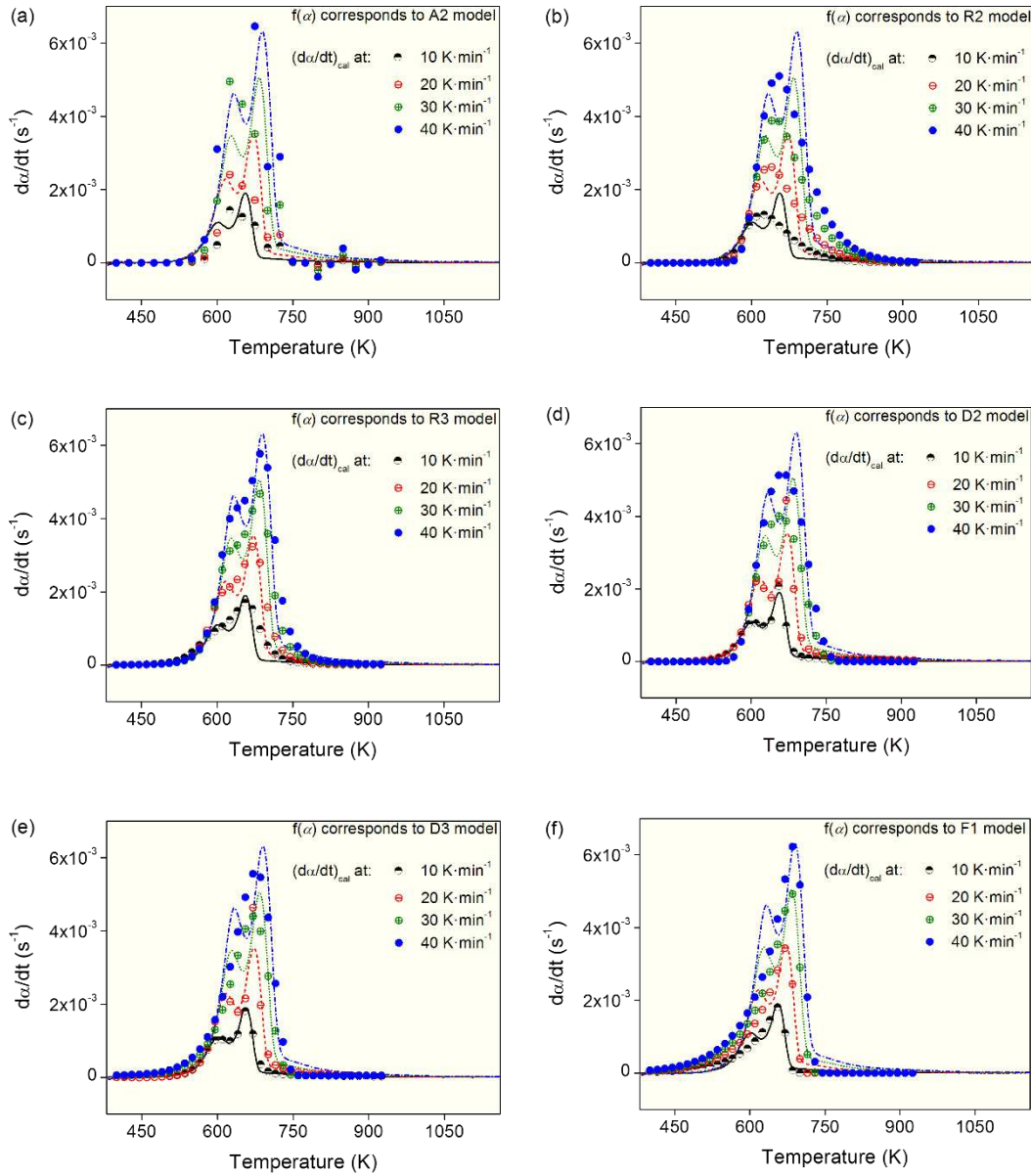
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Figure 5 (a - c) Flynn–Wall–Ozawa (FWO) plots and (d - f) Kissinger-Akahira-Sunose (KAS) plots for the thermal degradation reactions and conversion degrees of $\alpha = 0.05; 0.1; 0.2; 0.3; 0.4; 0.5; 0.6; 0.7; 0.8; \text{ and } 0.9$; for: (a and d) PSE 1: Hemicelluloses + extractives and lignin, (b and e) PSE 2: Cellulose + extractives and lignin, and (c and f) PSE 3: Lignin + extractives and residual holocellulose (hemicelluloses + cellulose)



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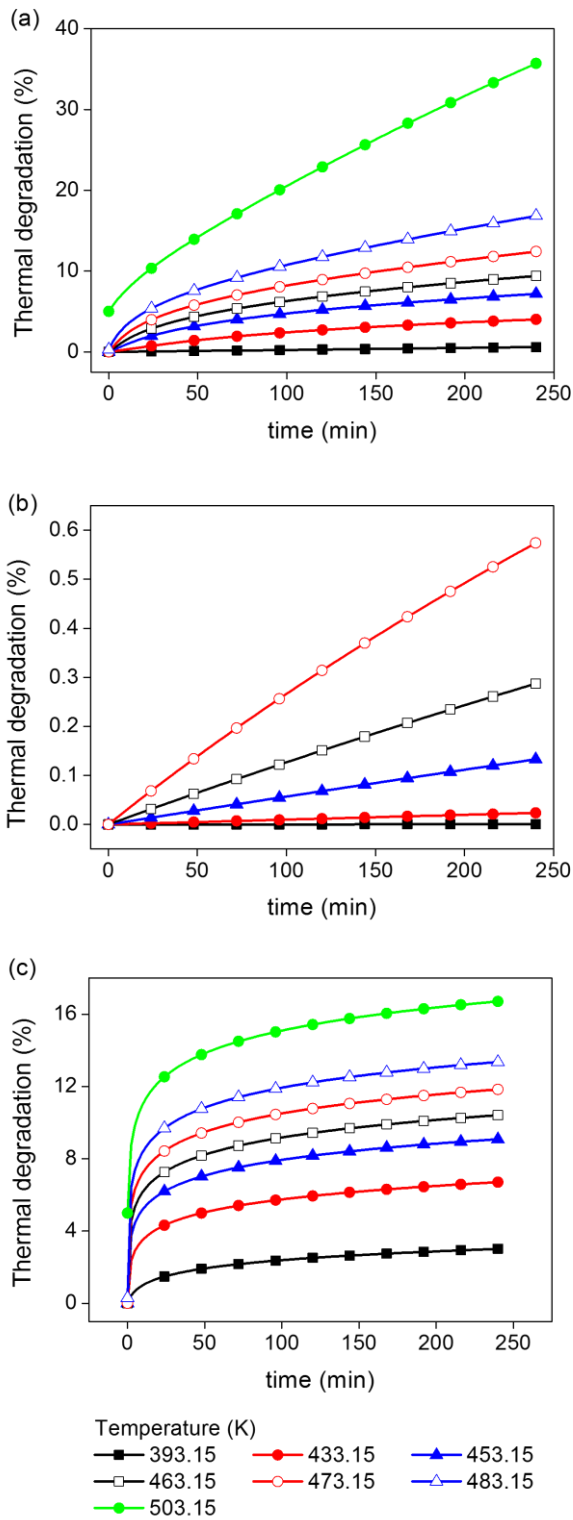
660 **Figure 6** $\left. \frac{d\alpha}{dt} \right|_{cal}$ best-fit of thermal degradation of SCB compared with $\left. \frac{d\alpha}{dt} \right|_{exp}$ curves at 10 K·min⁻¹; 20
 661 K·min⁻¹; 30 K·min⁻¹, and 40 K·min⁻¹ using different $f(\alpha)$ models used in the heterogeneous kinetic: (a) A2
 662 model; (b) R2 model; (c) R3 model; (d) D2 (model); (e) D3 model, and (f) F1 model



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665 **Figure 7** Thermal degradation fraction (TDF) corresponding to the parallel reactions scheme of (a) PSE 1:
 666 hemicelluloses + extractives and lignin, (b) PSE 2: cellulose + extractives and lignin, and (c) PSE 3: lignin+
 667 extractives and residual holocellulose (hemicelluloses + cellulose)



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