New insights on the kinetic analysis of isothermal data: the independence of the activation energy from the assumed kinetic model. *Pedro E. Sánchez-Jiménez*, Antonio Perejón, Luis A. Pérez-Maqueda, José M. Criado.* Instituto de Ciencia de Materiales de Sevilla, C.S.I.C.-Universidad de Sevilla, C. Américo Vespucio nº49, 41092 Sevilla, Spain

6

7 Abstract

8	Isothermal experiments are widely employed to study the kinetics of solid state reactions or
9	processes in order to extract essential kinetic information needed for modeling the processes at
10	an industrial scale. The kinetic analysis of isothermal data requires finding or assuming a
11	kinetic function that can properly fit the evolution of reaction rate with time so that the
12	resulting parameters, i.e. the activation energy and the preexponential factor, can be considered
13	reliable. In the present work, we demonstrate using both simulated and experimental data that
14	the kinetic analysis of a set of isothermal plots obtained at different temperatures, considering a
15	single step solid state reaction, necessarily leads to the real activation energy, regardless the
16	mathematical function selected for performing the kinetic analysis. This makes irrelevant the
17	election of the kinetic function used to fit the experimental data and greatly facilitates the
18	estimation of the activation energy for any single process.

19 Keywords: Kinetic Analysis, Isothermal; Kinetic Model; Activation Energy; Process

20 Modeling

^{*}Corresponding author. Tel +34954489548 Fax +34954460665 e-mail address: <u>pedro.enrique@icmse.csic.es</u>

22 **1. Introduction**

Kinetic analysis is widely employed as a tool for obtaining the essential knowledge needed for 23 modeling processes on an industrial scale. This is also true in the field of energy conversion 24 25 and production, with an important number of papers published every year in which the main objective is determining the kinetics governing processes such as pyrolysis, gasification, 26 combustion or thermal decomposition in order to optimize operating conditions ¹⁻¹⁴. The 27 28 experimental data are usually collected under isothermal or linear heating conditions. While linear heating rate experiments provide quickness and simplicity, in many studies such as those 29 involving long-term aging at operation temperatures ¹⁵⁻¹⁷, oxidation processes ¹⁸⁻²⁰, reaction 30 progress followed by spectral or DRX peak intensity measurements ²¹⁻²³, chemical looping 31 processes ^{1, 3, 11, 24} or those set-ups that try to replicate industrial operation conditions ^{9, 12, 13}, 32 isothermal experiments are still the most convenient or even feasible option. Moreover, 33 34 isothermal experiments present the distinct advantage of a higher capability for kinetic 35 mechanism discrimination due to the fact that the shape of the integral α -time curve is directly related to the obeyed model 25,26 . Thus, the α -time traces of phase boundary controlled reaction 36 (so called "n order" reactions) are convex, the diffusion controlled reactions are concave and 37 38 those whose rate is controlled by the formation and growth of nuclei (Avrami-Erofeev models) have sigmoidal shape. On the other hand, the α-temperature plots recorded under rising 39 temperature are always sigmoidal-shaped, whatever the reaction kinetic model ^{27, 28} 40

A former review ²⁹ on the kinetic dehydroxylation of kaolinite found that similar activation energies had been reported by different authors despite the proposal of different kinetic models. An analysis of those experimental data assuming a set of different kinetic models revealed that the activation energies obtained were independent of the kinetic model previously assumed, although no explanation was given. If such behavior was generalized, it would constitute an extraordinary advantage since it is generally assumed that the activation energy obtained by a

47	kinetic analysis is dependent on the kinetic law used to fit the experimental data ^{27, 30, 31} . Thus,
48	authors must spend a great deal of effort to determine the kinetic function most adequate to the
49	process under study in order to assure a reliable set of kinetic parameters ²⁷ . In the present work
50	we explore the influence the kinetic law selected to fit the experimental data has on the
51	activation energy yielded by the kinetic analysis of a set of isothermal curves recorded at
52	different temperatures. It is demonstrated first theoretically and then analyzing sets of both
53	simulated and experimental data, that the activation energy obtained by this kind of analysis
54	would always be the correct one, regardless the mathematical function selected for fitting the
55	data.

2. Experimental

Thermal degradation experiments were carried out using polytetrafluoroethylene (Aldrich,
product number 182478) at temperatures of 480, 490 and 500 °C in a Q5000IR TA Instruments
TGA (TA Instruments, Crawley, UK) connected to a gas flow system to work in inert
atmosphere equipment under 100 mL min⁻¹. Samples sizes of ~20 mg were placed in a
platinum crucible and heated at 300 °C min⁻¹ to the final temperature in order to avoid mass
loss before the steady state is attained.

3. Theory

It is well known that the kinetics of a solid state reaction, in conditions far from theequilibrium, can be described by the general equation:

$$\frac{d\alpha}{dt} = kf(\alpha) \tag{1},$$

where t represents the time, α is the extent of reaction and k is the rate constant, which depends 68 69 on the temperature according to the Arrhenius equation, $k=Aexp(-E_a/RT)$. The kinetic model, $f(\alpha)$, is a function describing the relationship between the reaction rate and the reacted fraction. 70 An ample selection of $f(\alpha)$ functions have been published along the last decades, ranging from 71 the widely employed first or nth order laws to more sophisticated diffusion or nucleation models 72 ³²⁻³⁴. It is important to remark that Eq. (1) does not consider any particular heating schedule so 73 74 it should be fulfilled whatever the time-temperature program employed for obtaining the 75 experimental data. Additionally, the reacted fraction or conversion α can be established using 76 data extracted with any technique measuring a property that can be directly related to the reaction rate, most usually the mass loss recorded by thermogravimetry. The standard 77 isothermal method of kinetic analysis follows a model-fitting approach²⁷. Thus, a series of 78 79 isotherms are recorded at different temperatures and the extracted experimental data are fitted 80 to a set of different kinetic models, according to the following equations, which are obtained by 81 integrating Eq (1):

82
$$\int_0^\alpha \frac{d\alpha}{f(\alpha)} = \int_0^t k dt$$
(2a)

83

$$g(\alpha) = kt \tag{2b},$$

where $g(\alpha)$ is the integral form of the kinetic model. The plot of $g(\alpha)$ versus the reaction time (provided that the time for reaching the temperature steady state is negligible with regards to the elapsed time) leads to a straight line whose slope is the rate constant, *k*. Then, given the Arrhenius dependence of the rate constant with the temperature, the activation energy can be subsequently calculated by plotting the logarithm of the rate constants versus the reverse of their corresponding temperatures:

$$\ln k = \ln A - E_a / RT \tag{3}$$

4

Let us assume the experimental kinetic data are fitted with a $G(\alpha)$ function different from the one really obeyed by the reaction, $g(\alpha)$, regardless of the quality of the resulting regression coefficient. In such case, an apparent rate constant, k_a , would be obtained from the plot of $G(\alpha)$ as a function of *t* according to the following equation:

95
$$G(\alpha) = k_a t \tag{4},$$

The comparison of Eqs (2b) and (4) leads to the conclusion that the acceptance of a reasonable linear correlation between $G(\alpha)$ and t necessarily implies to accept a linear correlation between $G(\alpha)$ and $g(\alpha)$, that would be expressed according to the following equation:

99
$$G(\alpha) = ag(\alpha) + b$$
 (5),

100 a and b being constants.

101 It can be concluded from Eqs. (2b), (4) and (5) that, whatever would be the temperature, the 102 apparent constant reaction rate, k_a is related with the actual one, k, through the following 103 relationships:

104
$$\frac{dG(\alpha)}{dt} = k_a = a \frac{dg(\alpha)}{dt} = ak$$
(6).

Eq. (6) shows that $k_a = ak$. Moreover, taking into account that k fits the Arrhenius equation, it follows:

$$k_a = ak = aAe^{-E/RT}$$
(7),

108 that could be rearranged in the following form:

109
$$\frac{-dlnk_a}{d(1/T)} = \frac{-dlnk}{d(1/T)} = E/R$$
 (8)

Thus, Eq. (8) clearly demonstrates that the activation energy determined from a set of
isothermal α-time plots obtained at different temperatures is independent of the kinetic model
previously assumed for performing the kinetic analysis.

113

114 4. Kinetic analysis of simulated isothermal curves

The time at which a given α value is reached at a certain temperature T can be determined,according to Eq. (2b), from the following expression:

117
$$t = \frac{g(\alpha)}{Ae^{-E/RT}}$$
(9)

Thus, Eq. (9) was used to numerically construct two sets of simulated curves assuming 118 119 temperatures of 250, 265, 275 and 300 °C. Figure 1a includes a set of isotherms simulated according to the following kinetic parameters: E=200 kJ/mol, A=10¹⁶ min⁻¹ and a first order 120 121 kinetic model (F1). On the other hand, isotherms in Figure 1b were constructed assuming the 122 same activation energy and pre-exponential factor as the previous set and an Avrami nucleation 123 kinetic model (A2). The figure clearly shows how the first order trace is convex while the 124 nucleation-driven trace is sigmoidal, as aforementioned. Then, the data from all the curves in Figure 1 were linearly fitted to several theoretical kinetic models, as per Eq (2b), producing the 125 126 rate constant values listed in Tables 1 and 2, together with their corresponding correlation 127 coefficients. In order to better replicate the analysis procedures most commonly employed in the literature, the fit was limited to data comprising the conversion range $0.1 \le \alpha \le 0.9$ since the 128 129 extreme ranges are more sensitive to experimental errors. Figure 2 includes a selection of plots 130 constructed using data from the 265 °C isothermal curve, providing a clear picture of the results 131 obtained. The linear fits from which slope the rate constants are determined are also marked in 132 the Figure. As expected, only the correct $g(\alpha)$ function produces a flawless linear fit whereas

133	the fit to incorrect models yield non-linear plots with different slopes and, therefore, lead to
134	different rate constants. Nevertheless, it is worth mentioning that 5 out of the 8 models tested
135	yielded significant correlation coefficients, over 0.99. Even more striking are the conclusions
136	reached when the activation energy is determined for each set of rate constants (Tables 1 and 2)
137	as per Eq. (3). Table 3 lists the activation energies and pre-exponential factors obtained,
138	depending on the model used to construct the simulated curves and the model used to fit the
139	data. Results are identical for every case, with the analysis yielding the correct activation
140	energy regardless the model assumed. Thus, the erroneous selection of a kinetic law, and
141	consequently, the erroneous estimation of the rate constants have no influence whatsoever on
142	the obtained activation energy. On the other hand, the pre-exponential factors present a slight
143	variability as it is expected from Eq. (7), although it is still well within the accepted error range
144	given the high numerical value of such constant. It should be warned that any prediction
145	attempt requires the knowledge of correct kinetic model driving the proccess. Thus, for a given
146	temperature, each kinetic triplet in Table 3 will produce a different α -time curve, with only the
147	right model been able to accurately predict the experimental curve. Nevertheless, it is still
148	possible to make reliable predictions from the model-independent E values obtained by the
149	isothermal method by employing Vyazovkin's isoconversional equation. ³⁵

151 **5.** Kinetic analysis of experimental isothermal curves

Next, for further confirmation, real experimental data are tested. Figure 3 includes three isothermal curves, corresponding to the decomposition of polytetrafluoroethylene (Aldrich, product number 182478) recorded at temperatures of 480, 490 and 500 °C in a Q5000IR TA Instruments TGA equipment under a 100 mL min⁻¹ flow of N₂. The plots of $g(\alpha)$ versus the reaction time, as per Eq (2b), for each isothermal curve were built using eight different kinetic models. A selection of these plots is shown as examples in Figure 4. The rate constants, as

158	directly calculated from the slope of the plots, are included in Table 4, as well as the activation
159	energies determined using each set of rate constants, as previously described. Additionally, the
160	plots of the rate constants versus the reverse of the temperature (as per Eq. (3)), from which the
161	activation energies are calculated, are included in Figure 5. These results indicate that, as
162	predicted by the mathematical development presented in a previous section, the same value for
163	the activation energy, 261 ± 1 kJ/mol, is reached whatever the kinetic model used to fit the data.
164	The activation energy here obtained is in agreement with that estimated in a previous study
165	using the same material and employing a combined approach based on bothmodel-fitting and
166	isoconversional methods, thus confirming the validity of the results. ³⁶ Additionally, as it
167	happened with the simulated curves, four out of eight models can fit the experimental data with
168	reasonable correlation coefficients.

This finding entails significant implications. The fact that the correct activation energy of a 169 170 single process would always be obtained from a set of isotherms regardless of the kinetic model 171 chosen to fit the data permits the isothermal method of kinetic analysis to provide the activation 172 energy of any single step reaction without needing any previous knowledge regarding the 173 reaction mechanism. In any case, heat and mass transfer limitations typical of isothermal 174 experiments must still be considered in the experiments design since they will inevitably 175 produce interferences with the real process if not adequately minimized. The ability of the 176 method to yield the correct activation energy regardless of the model used is especially interesting if we consider that most reactions will rarely follow faithfully any theoretical 177 models, which were built upon several ideal assumptions and constraints which are seldom 178 179 fulfilled in real reactions. For instance, inhomogenous distribution in size and particle shape have been shown to have an important effect on the shape of the experimental curves.³⁷ Such 180 181 deviations would have consequences in model-fitting methods of kinetic analysis of linear 182 heating rate experiments because the activation energy and the pre-exponential factors provided are heavily dependent on the kinetic model used for the fitting. ^{31, 38, 39} On the other hand, as
shown here, the activation energy and pre-exponential factors provided by the isothermal
method are independent of the model used to fit the experimental data, hence preventing the
distorting effect of non-ideal models or inhomogenous materials.

187

188 6. Conclusions

It has been proven that the actual activation energy of any single step solid state reaction can 189 190 be determined from a set of isothermal experiments regardless of the kinetic model of function obeyed by the reaction and/or the kinetic equation previously assumed for performing the 191 192 kinetic analysis. Thus, the isothermal method of kinetic analysis behaves at all effects as a 193 model-free since the activation energy and the pre-exponential factor can be calculated without 194 a previous knowledge of the kinetic model. Moreover, the kinetic parameters thus obtained 195 would be representative of the reaction even when none of the theoretical models could closely 196 represent the studied process.

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264	Figure Captions
265	Figure 1. Isothermal curves simulated according to the following kinetic parameters:
266	E=200kJ/mol, A=10 ¹⁶ min ⁻¹ and (a) F1 kinetic model, $f(\alpha)=(1-\alpha)$ or (b) A2 Avrami model,
267	$f(\alpha)=2(1-\alpha)(-\ln(1-\alpha))^{0.5}.$
268	Figure 2. Fit of the 265°C isothermal curves simulated assuming (a) F1 and (b) A2 models to a
269	set of different kinetic models.
270	Figure 3. Isothermal curves corresponding to the degradation of PTFE, recorded at 480, 490
271	and 500°C.
272	Figure 4. Fit of the PTFE degradation isothermal curves in Figure 3 to four different kinetic
273	models: first order (F1), phase boundary controlled (R2), two-dimnsional diffusion (D2) and
274	nucleation (A2).

- Figure 5. Plot of lnk versus the reverse of their corresponding temperature, constructed for
- every set of rate constants in Table 4. Activation energy is extracted from the slope of the plots,
- 277 being identical in every case.



283	Figure 1. Isothermal curves simulated according to the following
284	kinetic parameters: E=200kJ/mol, A= 10^{16} min ⁻¹ and (a) F1 kinetic
285	model, $f(\alpha)=(1-\alpha)$ and (b) A2 Avrami model, $f(\alpha)=2(1-\alpha)(-\ln(1-\alpha))^{0.5}$.





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 Figure 2.

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 (a) F1 and

Figure 2. Fit of the 265°C isothermal curves simulated assuming (a) F1 and (b) A2 models to a set of different kinetic models.



PTFE, recorded at 480, 490 and 500°C.

Figure 3. Isothermal curves corresponding to the degradation of



Figure 4. Fit of the PTFE degradation isothermal curves in Figure 3 to four
 different kinetic models: first order (F1), pase boundary controlled (R2), two dimensional diffusion (D2) and nucleation (A2).





320 Table 1. Rate constants and correlation factors obtained from fitting simulated curves in Figure

321 1a (assuming a F1 model) to different kinetic models, according to Eq. (2a).

2	n	2
3	2	2

	250°C		265°C		275°C		300°C	
Model Fitted to	k (min ⁻¹)	r ²						
F1	0.0062	1.000	0.0224	1.000	0.0507	1.000	0.3441	1.000
R2	0.0038	0.981	0.0136	0.980	0.0309	0.980	0.2097	0.980
R3	0.0044	0.991	0.0160	0.991	0.0362	0.991	0.2456	0.991
A2	0.0034	0.970	0.0122	0.970	0.0275	0.970	0.1868	0.970
A3	0.0024	0.944	0.0085	0.944	0.0192	0.944	0.1305	0.944
D2	0.0020	0.996	0.0073	0.996	0.0165	0.996	0.1124	0.996
D4	0.0005	0.993	0.0020	0.993	0.0044	0.993	0.0300	0.993
L2	0.0145	0.996	0.0524	0.996	0.1186	0.996	0.8060	0.996

Table 2. Rate constants and correlation factors obtained from fitting the simulated curves in

Figure 1b (assuming a A2 model) to different kinetic models, according to Eq. (2a).

	250°C		265°C		275°C		300°C	
Model Fitted to	k (min ⁻¹)	r ²						
F1	0.0111	0.970	0.0400	0.970	0.0905	0.970	0.6147	0.970
R2	0.0069	0.997	0.0250	0.997	0.0565	0.997	0.3836	0.997
R3	0.0080	0.992	0.0290	0.992	0.0656	0.992	0.4458	0.992
A2	0.0062	1.000	0.0224	1.000	0.0507	1.000	0.3441	1.000
A3	0.0044	0.996	0.0158	0.996	0.0358	0.996	0.2430	0.996
D2	0.0036	0.953	0.0130	0.953	0.0293	0.953	0.1994	0.953
D4	0.0010	0.936	0.0034	0.936	0.0078	0.936	0.0529	0.936
L2	0.0263	0.988	0.0947	0.988	0.2143	0.988	1.4563	0.988

332 Table 3. Activation energies, pre-exponential factors and correlation coefficients obtained from

- the plot of the different rate constants in Tables 1 and 2 versus the reverse of their
- 334 corresponding temperature, as per Eq.(3).

Model used to simulate the curves F1 A2									
Model fitted to	E (kJ mol ⁻¹)	A (min ⁻¹)	r ²	E (kJ mol ⁻¹)	A (min ⁻¹)	\mathbf{r}^2			
F1	200	5×10^{16}	1.000	200	1×10^{17}	1.000			
R2	200	5×10^{16}	1.000	200	6×10^{16}	1.000			
R3	200	4×10^{16}	1.000	200	8×10^{16}	1.000			
A2	200	$3x10^{16}$	1.000	200	6×10^{16}	1.000			
A3	200	$2x10^{16}$	1.000	200	$4x10^{16}$	1.000			
D2	200	$2x10^{16}$	1.000	200	$4x10^{16}$	1.000			
D4	200	5×10^{15}	1.000	200	1×10^{16}	1.000			
L2	200	1×10^{17}	1.000	200	$3x10^{17}$	1.000			

³³⁵

Table 4. Rate constants and correlation factors calculated from the fit of the experimental

338 curves corresponding to the degradation of PTFE (Figure 3) to a set of theoretical kinetic

models. The activation energy and correlation coefficients obtained from the plot of the rate

340 constants versus the reverse of the temperature (Eq.(3)) are also included.

	480°C		490°C		500°C		Result of Isothermal analysis	
Model Fitted to	k (min ⁻¹)	r ²	k (min ⁻¹)	r ²	k (min ⁻¹)	r ²	E (kJ mol ⁻¹)	r ²
F1	0.0047	1.000	0.0092	1.000	0.0164	0.999	261±10	0.997
R2	0.0026	0.978	0.0052	0.979	0.0093	0.982	261±11	0.996
R3	0.0032	0.990	0.0062	0.990	0.0111	0.993	261±11	0.996
A2	0.0023	0.973	0.0046	0.974	0.0082	0.977	260±11	0.996
A3	0.0016	0.951	0.0031	0.952	0.0055	0.955	261±12	0.996
D2	0.0015	0.997	0.0030	0.998	0.0054	0.998	262±11	0.996
D4	0.0004	0.997	0.0008	0.997	0.0015	0.997	260±10	0.997
L2	0.0106	0.996	0.0208	0.997	0.0373	0.997	261±10	0.996

³³⁶