



Thermogravimetric Characteristics and Non-isothermal Kinetics of Macro-Algae With an Emphasis on the Possible Partial Gasification at Higher Temperatures

Imtiaz Ali* and Ali Bahadar

Department of Chemical and Materials Engineering, King Abdulaziz University, Rabigh, Saudi Arabia

Pyrolysis of *Turbinaria ornata* was realized in a thermogravimetric analyzer. The process was crudely classified into primary and secondary reaction zones. In the primary reaction zone, the thermal decompositions of the low-thermal stable components produced volatiles and biochar. The solid products obtained from the primary decomposition reactions contained inorganics with heavy metals. At mild-to-high temperatures, the catalytic effects accompanied gasification using oxygen, which was partially supplied by the oxygen carriers present in the solids and evolved gases. In order to study the pyrolytic conversion, combined, and multiple reaction schemes were employed. While the model-free methods helped to provide the accurate activation energies and the initial value of the pre-exponential factor, the non-linear regression optimized the chosen model parameters. In this study, a simple order-based model was compared with the versatile Šesták-Berggren (SB) model considering combined and multiple reactions. The application of a multiple reaction scheme to the primary and secondary reaction zones concluded that a simple order-based model suffices for the kinetic analysis. The secondary decomposition was shown to start with a high activation energy, which decreased appreciably when the conversion proceeded toward completion.

OPEN ACCESS

Edited by:

Mohammad Rehan, King Abdulaziz University, Saudi Arabia

Reviewed by:

Meisam Tabatabaei, Agricultural Biotechnology Research Institute of Iran, Iran Vivekanand Vivekanand, Malaviya National Institute of Technology, Jaipur, India

> *Correspondence: Imtiaz Ali imtiaz_che@hotmail.com

Specialty section:

This article was submitted to Bioenergy and Biofuels, a section of the journal Frontiers in Energy Research

Received: 19 October 2018 Accepted: 18 January 2019 Published: 13 February 2019

Citation:

Ali I and Bahadar A (2019) Thermogravimetric Characteristics and Non-isothermal Kinetics of Macro-Algae With an Emphasis on the Possible Partial Gasification at Higher Temperatures. Front. Energy Res. 7:7. doi: 10.3389/fenrg.2019.00007 Keywords: Turbinaria ornata, seaweed, pyrolysis, pseudo-components, independent parallel reactions

INTRODUCTION

The last few decades have seen an increase in the use of biofuels and high-value biochemicals derived from biomass. Biomass is a renewable carbonaceous source, an alternative to fossil fuel, and a precursor of biochemicals. Non-food biomasses are widespread in nature and are also cultivated at ~100 billion tons annually (Sheldon, 2014). According to an estimate, biofuels are expected to satisfy around 10% of the global energy demand by 2035, and they have the potential to replace 27% of global transportation fuel by 2050. The main reason for the growing interest in the use of biomass, besides its renewable nature, is its carbon-neutral and less polluting characteristics. The Paris climate agreement requires clean energy transformation in order to reach net zero emissions by 2060, which can only be achieved by scaling-up efficient energy conversion systems to generate bio-energy with carbon capture and storage (BECCS) (Agency, 2017).

Marine macro-algae or seaweeds are efficient photosynthetic organisms. Besides their use in the food, pharmaceutical, and cosmetic industries, they are also well-known for their use in bio-sequestration and bioremediation, thereby mitigating climate change (Maceiras et al., 2011; Oliveira, 2016). There has been increased interest in macro-algae as a renewable and sustainable energy source over the past decade among the scientific and industrial communities because of its high biomass productivity (Baghel et al., 2017). Macro-algae biomass is mainly produced from the harvesting of wild stocks, a recent report of FAO reveals that a volume of around 12 million tons is produced annually (Ferdouse et al., 2018). Its cultivation is normally carried out through ocean farming and thus does not require land or fresh water, allowing these to be used for food crops and aquaculture. Macro-algae is a cheap source of carbon with a varying chemical composition depending on the type and the environmental conditions. It can be crudely classified into four main groups depending on the stored nutrients, pigments and chemical composition, namely red seaweeds (Rhodophyceae), green seaweeds (Chlorophyceae), brown seaweeds (Phaeophyceae), and blue-green seaweeds (Cyanophyceae) (Manivannan et al., 2009).

Seaweeds, like terrestrial forms of biomass, are lignocellulosic in nature with a complicated embedding of cellulose, hemicellulose, lignin, and intracellular substances. However, compared to terrestrial biomass, they may have higher costs attached to their cultivation in conjunction with the presence and effects of heavy metal concentrations in aquatic environments, while the removal of harmful contents (sulfur and nitrogen) from the resulting fuel (Ghadiryanfar et al., 2016) is another problem. Furthermore, some studies have revealed that the transesterification of oil extracted from macro-algae is restricted due to low lipid contents, and they have suggested the conversion of carbohydrates into biofuel but with a higher gas yield (Roesijadi et al., 2010; Hong et al., 2017). Therefore, the most suited conversion technologies, such as anaerobic digestion and hydrous pyrolysis (Ross et al., 2008), are likely to be employed for macro-algal biomass containing high levels of ash and alkali metals. Macro-algae are rich in macro and micronutrients, but also display an uptake of heavy metals from the environment. The heavy metals in macro-algae may indicate the fractions of industrial effluents affecting the aquatic ecosystems, which is also a serious concern. These accumulations depend upon their biochemical composition, specifically the properties of the cell wall (Davis et al., 2003) and the constituents present in the macro-algae cell walls. The cell wall contains the fibers and matrix of polysaccharides (alginates and fucoidan). The alginates in brown algae have a high accumulation of divalent cations, while sulfated polysaccharides exhibit a tendency toward trivalent cations (Saha and Orvig, 2010; Ortiz-Calderon et al., 2017). Figure 1 presents the various major constituents of the seaweed cell wall.

The uptake of heavy metals takes place in macro-algae through bio-sorption or ion exchange processes by the chemical constituents present in the cell wall. The Red Sea, with its diverse ecosystem, has over 500 species of seaweeds, which are spread over 1,900 km covering 438,000 km² of coral reef



stretching along its coasts and part of the Indian-Pacific Ocean (Spalding et al., 2001; Siddall et al., 2004; Bruckner et al., 2013). Red Sea seaweeds primarily consist of small green and brown algal filaments found in the northern and central parts, while the southern area of the Red Sea has large brown algal species like Sargassum and Turbinaria spp (Bruckner et al., 2013). Turbinaria ornata (T. ornata) is a well-known brown alga suitable for human consumption due to its antioxidant characteristics (Deepak et al., 2017a). Several studies have discussed its antibacterial, antioxidant, and anticancer properties (Omar et al., 2012; Ponnan et al., 2017; Stranska-Zachariasova et al., 2017; Tenorio-Rodriguez et al., 2017) as well as the biogenic synthesis of crystalline Ag/Au nanoparticles (Rajeshkumar et al., 2013; Kayalvizhi et al., 2014; Deepak et al., 2017b), but to the best of the authors' knowledge, no previous studies are available for its use as a source of biofuel. The mineral and heavy metal uptakes by different *Turbinaria* species are presented in Table 1.

Chemical looping pyrolysis gasification (CLPG) and chemical looping combustion (CLC) are proven concepts for syngas production and CO₂-capturing, respectively. The use of oxygen carriers such as the oxides of Ni, Mn, Cu, Fe, and Co have previously been investigated in these processes (Adanez et al., 2012; Huang et al., 2017). In the present study, the pyrolysis of T. ornata was realized in a thermogravimetric analyzer (TGA). The unique and novel concept of pyrolysis-gasification with the oxides of heavy metals already embedded in the matrix of polysaccharides of brown macro-algae (T. ornata) has also been considered besides the condensation of polymers at higher temperatures. The understanding of the role of inherent inorganic compounds of *T. ornata* and the investigation into evolved gases at high temperatures will lead to process optimization, efficient use, and the removal of heavy metals in the form of biochar.

MATERIALS AND METHODS

Macro-Algae Biomass

T. ornata is a brown alga that has phytochemical constituents like carbohydrates, proteins, flavonoids, alkaloids, and phenols as well as sulfated fucan-polymers like polysaccharide and alginate

Macroalgae species		Micro/Macro	o minerals				Í	eavy met	tals					References
	×	Ca	Na	Mg	Ъе	Zn	Mn	As	ଞ	ŗ	Hg	ïŻ	Pb	
Turbinaria omata	112.2 ± 5.7	41.3 ± 1.2	19.3 ± 0.5	12.7 ± 0.5	0.891 ± 0.211	0.042 ± 0.007	0.015 ± 0.007	1	I	1	1	1	I	Zubia et al., 2005
(uo I inear I on)	35.2 ± 0.7	29.40 ± 0.15	36.10 ± 0.16		0.49 ± 0.01	0.69 ± 0.01	0.58 ± 0.01	I	I	I	I	I	I	Mohamed Ali and Elnabawy Ward, 2016
Turbinaria. conoides mg g^{-1} dw (mean \pm SD)	27.9 ± 1.1	14.8 ± 2.2	11.5 ± 0.5	5.7 ± 0.3	0.062 ± 0.017	0.006 ± 0.003	1	1	1	1	1	1	I	Santoso et al., 2006
Turbinaria triquetra (ppm)	I	I	1	1	I	I	1	7.79	0.06		4.65	0.55	0.89	El-Naggar and Al-Amoudi, 1989
<i>Turbinaria triquetra</i> (mg kg ⁻¹ , dry weight)	I	1	1	1	I	I	1	-	0.80 ± 0.3	0.10 6	·2 ± 2.4	1.1 ± 0.2 (3.1 ± 1.1	Abdallah et al., 2006
Turbinaria deccurens (ppb)	1	1	1	1	I	I	1	99.53	23.47	5.71	1	1	1.47	Roy et al., 2017
Turbinaria omata (ppb)	1	1	1	1	I	I	1	129.37	3.07	2.04	1	1	1.92	Roy et al., 2017

structures that display tremendous potential for use in various applications. Its ability to bind heavy metals is of particular interest. Its biochar can also be used as a bio-indicator for identifying the type and quantity of heavy metals present in aquatic environments.

T. ornata was collected from the Red Sea coastal area $(N22^{\circ}48'32'', E38^{\circ}56'37'')$. The fresh macro-algae was thoroughly washed to remove adherent dirt and sand with deionized water several times. The cleaned biomass was air-dried for 2 days, then crushed and sieved. A fine powder with a particle size of <0.147 mm (the fraction of the material passed through a 100 mesh sieve) was kept in an air-dried container for further use. **Table 2** shows the physicochemical properties of *T. ornata*.

Thermogravimetric Analysis

Pyrolysis is a promising process to convert biomass whereby the distribution of the product depends greatly on the conditions used. It has been reported that slow pyrolysis yields more biochar whereas fast pyrolysis produces more volatiles (Mohan et al., 2006). Extensive development in the pyrolysis process has been made recently and there has been a growing interest in its use because of its simplicity, flexibility, and efficiency. Biochar, a product of pyrolysis, finds in addition to volatiles potential application in the agricultural and environmental sectors. A thermogravimetric analyzer (TGA) is a commonly employed thermo-analytic instrument for the thermal degradation study of biomass. Under well-defined conditions of slow pyrolysis, it provides precise time and temperature-series mass loss data, which are used to estimate the kinetic parameters.

In the present study, the TGA experiments were performed on a Q5000 IR electro-balance analyzer (TA Instruments, USA). Approximately 15 mg of algal biomass sample was used for each run in triplicate to ensure accuracy and precision. The TGA temperature was raised from 50 to 900°C using different heating rates (5, 10, 15, and 30 °C·min⁻¹), while a flow of high purity N₂ gas was maintained at 70 ml min⁻¹. Additionally, an isothermal step was introduced before the heating ramp to achieve pyrolysis stabilization. The acquisition, diagnosis, and processing of curves were performed on PlatinumTM software.

Kinetic Evaluation and Analysis

The devolatilization of algal biomass through pyrolysis is useful in understanding multi-component and multi-pyrolytic phases for process design, optimization, and scale-up (Ranzi et al., 2017).

TABLE 2	Physicochemical	properties	of	Turbinaria	ornata.
---------	-----------------	------------	----	------------	---------

Proximate a	nalysis ^a	Ultim	ate analysis ^b
Moisture	15.6%	С	30.38%
Volatile Matter	54.2%	Н	6.03%
Ash	25.3%	Ν	0.89%
Fixed Carbon	4.9%		
Ash Fixed Carbon	25.3% 4.9%	Ν	0.89%

^aASTM standard methods (E1756, E1755-01, and D3174-12) were used except for fixed carbon, which was calculated from the difference.

 b C, H, and N were measured with a CHN/O 2400 series II Perkin Elmer elemental analyzer.

TABLE 1 | Macro/micro mineral and heavy metal uptake by Turbinaria macroalgal species

A global single-step reaction is always used in assessing the kinetic mechanism of pyrolysis. In a simplistic approach, the overall reaction process of *T. ornata* pyrolysis can be described with the following reaction:

Turbinaria ornata
$$(s) \rightarrow Volatiles (g) + Biochar (s)$$
 (1)

A non-isothermal TGA analysis technique is usually preferred over an isothermal technique for three main reasons. Firstly, the isothermal procedures are simply not possible to realize at higher temperatures because of the large non-isothermal heat-up and cool-down durations. Secondly, the estimation of the kinetic parameters at low heating rates is rather difficult due to weight loss during the heat-up time. Thirdly, there is the existence of the almost non-zero extent of conversions (Vyazovkin and Wight, 1998; Grigiante et al., 2016). For the aforementioned reasons, a non-isothermal technique was selected for this study. The process data recorded by TGA during non-isothermal pyrolysis can be expressed in terms of conversion (α) at any time (t) using the Arrhenius law [43]:

$$\frac{\mathrm{d}\alpha}{\mathrm{d}t} = \beta \frac{\mathrm{d}\alpha}{\mathrm{d}T} = A \cdot \exp\left(-\frac{E_a}{RT}\right) \cdot f\left(\alpha\right) \tag{2}$$

where E_a (J·mol⁻¹) is the apparent activation energy, A (s⁻¹) is the frequency factor, T is the absolute temperature, R is the ideal gas constant (J·mol⁻¹·K⁻¹), β is the heating rate (K·s⁻¹), and $f(\alpha)$ represents the algebraic form of the reaction model. The conversion (α) is defined as the fractional mass loss expressed by the following equation:

$$\alpha = \frac{m_o - m_i}{m_o - m_f} \tag{3}$$

where m_0 , m_i , and m_f are the initial, instantaneous, and final mass, respectively.

Integrating Equation (1) with respect to conversion and temperature in the ranges $\alpha \in [0, \alpha]$ and $T \in [T_o, T]$ gives Equation (4):

$$g(\alpha) = \int_{0}^{\alpha} \frac{\mathrm{d}\alpha}{f(\alpha)} = \frac{A}{\beta} \int_{T_0}^{T} \exp\left(-\frac{E_a}{RT}\right) \mathrm{d}T$$
(4)

Equation (4) can be used to describe the conversion when the temperature lag between the sample and its environment is very small. E_a is an important kinetic parameter, defined as the energy required to have a reaction, which is usually measured by model-free and model-fitting methods. Each method has its own advantages, and they are complementary and are not in competition. A model-free analysis can be performed without having specified the reaction mechanism, while the model-fitting analysis uses the approach of data-model best fitting to estimate E_a , A, and $f(\alpha)$ (Šesták and Berggren, 1971).

Model-Free Methods

Model-free kinetics methods are mostly isoconversional in nature. They provide an accurate estimation of E_a from

isothermal and non-isothermal measurements and are most commonly employed in the pyrolytic kinetics of biomass (Opfermann et al., 2002). These methods have the capability to solve complex processes and are based on the fact that the reaction rate at a particular heating rate (β) is a function of temperature (T) alone at a specific conversion (α), while $f(\alpha)$ remains constant. Model-free methods provide a detailed kinetic analysis without knowledge of the reaction mechanism while indicating multiple processes if E_a varies strongly with α (Vyazovkin et al., 2011). Depending on the mathematical form of the kinetic equation, model-free methods can be differential or integral. The Friedman method (Friedman, 1964) is the most common differential while the Kissinger-Akahira-Sunose (KAS) method (Kissinger, 1957) is a widely used integral method. The Friedman method reveals the dependency of the conversion rate $\ln\left(\frac{d\alpha}{dt}\right)_{\alpha}$ over $1/T_{\alpha i}$ as a straight line:

Friedman:
$$\ln\left(\frac{\mathrm{d}\alpha}{\mathrm{d}t}\right)_{\alpha_i} = \mathrm{const} - \frac{E_\alpha}{R \cdot T_{\alpha_i}}$$
 (5)

The apparent activation energy (E_{α}) at a particular conversion (α) is estimated from the slope of the straight line. Integral methods are relatively less sensitive to data noise as compared to differential isoconversional methods and offer consistent E_{α} measurements. The KAS method is represented by the following linear Equation (6):

KAS:
$$\ln\left(\frac{\beta_i}{T_{\alpha_i}^2}\right) = \operatorname{const} - \frac{E_{\alpha}}{R \cdot T_{\alpha_i}}$$
 (6)

The slope of the linear fit of the $\ln\left(\frac{\beta_i}{T_{\alpha_i}^2}\right)$ vs. $1/T_{\alpha i}$ is used to evaluate the apparent E_{α} .

Model-Fitting Methods

The model-fitting analysis is conventionally used for the pyrolytic assessment of biomass materials on the basis of their devolatilization weight loss measurements using 1storder decomposition single-step reaction modeling or adding the weight of multiple pseudo-components like cellulose, hemicellulose, lignin, and others (Aboyade et al., 2012). The resulting characteristics from the kinetic analysis by model-fitting yield detailed simulations of biomass pyrolysis experiment data. Also, many studies have reported more precise simulations when using nth-order for specific reaction mechanisms where an optimization of the unknown constant is needed (Osman et al., 2017). Thus, in model-fitting procedures many models were applied to data collected from experiments to achieve the best fit to assess the kinetic parameters.

Combined Kinetics

The simple nth-order reaction has been used extensively to study the pyrolytic kinetics of biomass materials (Xu et al., 2017). The nth-order reaction mechanism for a solid state generally uses an expression like $f(\alpha) = (1 - \alpha)^n$, which when used in Equation (2) becomes:

$$\frac{\mathrm{d}\alpha}{\mathrm{d}t} = A \cdot \exp\left(-\frac{E_a}{RT}\right) \cdot (1-\alpha)^n \tag{7}$$

Another function is the Šesták—Berggren model (Šesták and Berggren, 1971), which was employed for the kinetic fitting of the experimental data. This is because of the high flexibility of the function to investigate the various types of physicogeometrical mechanism of the solid-state reaction (Kissinger, 1957; Opfermann et al., 2002; Vyazovkin et al., 2011; Aboyade et al., 2012), deviating cases with non-integral or fractal reaction geometry (Šimon, 2011; Dussan et al., 2017; Osman et al., 2017), the size distribution of the reactant particles (Orfão et al., 1999), and so on (Zubia et al., 2005). The high flexibility makes it difficult to interpret the physicochemical meanings of the kinetic exponent, but the SB model with the three parameters indicates the highest performance as a fitting function.

SB:
$$f(\alpha) = \alpha^m \cdot (1-\alpha)^n \cdot \left[-\ln(1-\alpha)\right]^p$$
 (8)

The SB model with three fit parameters (m, n, and p) is considered as one of the highest performing model-fitting methods, where *m* is accelerating behavior, *n* represents the order of the reaction and *p* shows the nuclei growth. By incorporating the SB model, Equation (7) becomes:

$$\frac{\mathrm{d}\alpha}{\mathrm{d}t} = A \cdot \exp\left(-\frac{E_a}{RT}\right) \cdot (1-\alpha)^n \cdot \alpha^m \cdot \left[-\ln\left(1-\alpha\right)\right]^p \quad (9)$$

The SB (*m*, *n*, *p*), for specific reaction model functions and other parameters like E_a and *A* can be optimized by employing a nonlinear least square analysis, whereby the residual sum of squares, also known as the sum of squared errors (SSE), is minimized by fitting the data curve of $\frac{d\alpha}{dt}\Big|_{exp}$ from the predicted curve of $\frac{d\alpha}{dt}\Big|_{exp}$ against time.

$$SSE = \sum \left(\left. \frac{\mathrm{d}\alpha}{\mathrm{d}t} \right|_{exp} - \left. \frac{\mathrm{d}\alpha}{\mathrm{d}t} \right|_{pred} \right)^2 \tag{10}$$

A non-linear least square analysis always requires the initial values for SB (m, n, p), E_a and A for optimization. After fitting

the data, the goodness of the fit was measured by *Fit* (%), as in Equation (11).

$$Fit (\%) = \left[1 - \frac{\sqrt{SSE}}{\left(\frac{d\alpha}{dt}\Big|_{exp}\right)_M}\right] \times 100$$
(11)

where $\left(\left. \frac{d\alpha}{dt} \right|_{exp} \right)_M$ is the peak-maximum conversion rate.

Independent-Parallel Reactions

Independent reaction models are mainly employed to identify the roles of the components in the thermal decomposition process. Biomass consists of many components and each component behaves differently. Therefore, these models have a set of discrete multiple reactions and may be extended to continuous distribution, which is largely employed to study the degradation kinetics of lignocellulosic materials. It is commonly accepted that each pseudo-constituent degrades independently at a particular temperature range. The weight loss calculations are performed by taking the sum of the mass fractions from the individual pseudo-components' decomposition rates (Orfão et al., 1999). The polysaccharides of *T. ornata* are mainly hemicellulose, cellulose, and others. The proximate composition of different *Turbinaria* species is given in **Table 3** below.

T. ornata contains many components and each component can be divided into subcomponents. During decomposition, the components which degrade during a similar temperature range may be grouped together and termed as pseudo-component. Lignocellulosic materials are normally classified into three pseudo-components, whereby each pseudo-component is involved in an independent reaction (Orfão et al., 1999). Similarly, the pyrolysis of *T. ornata* can be presented in terms of three pseudo-independent components, as given in Equation (12).

$$\begin{array}{c} Pseudo - Hemicellulose \xrightarrow{\frac{d\alpha_1}{dt}} \\ Pseudo - Cellulose \xrightarrow{\frac{d\alpha_2}{dt}} \\ Others \xrightarrow{\frac{d\alpha_3}{dt}} \end{array} \end{array} \} Volatiles1 + Biochar1 \quad (12)$$

During the main stage of the pyrolysis process, the pseudocomponents are subjected to the thermal splitting of chemical

TABLE 3 Proximate cor	nposition of various Turk	binaria species.					
Composition	Ash	Total fiber	Crude lipid	Crude protein	Fat	Moisture	References
Turbinaria ornata % dw (mean ± SD)	39.8 ± 0.8	38.1±1.5	2.2 ± 1.1	9.2 ± 0.6	_	10.8 ± 3.9	Behairy and El-Sayed, 1983
Turbinaria conoides % dw (mean ± SD)	2.5 ± 0.1	9.5 ± 0.5	-	1.0 ± 0.1	0.8 ± 0.1	85.1 ± 0.3	Mohammadi et al., 2013
Turbinaria murrayana % dw	17.04	40.06	1.46	19.51	-	-	Santoso et al., 2006
Turbinaria triquetra % dw (mean \pm SD)	29.573 ± 0.007	48.3 ± 0.007	1.62 ± 0.003	4.133 ± 0.009	-	23.62 ± 0.42	Zubia et al., 2005

TABLE 3 | Proximate composition of various Turbinaria species

bonds for the production of primary Volatiles1, gases (e.g., CH₄, CO2, and CO), and condensates and yield biochar solids at <500°C, named as Biochar1. Biochar contains minerals and even trace elements which were present in parent biomass. The pyrolytic conversion of three pseudo-components (i.e., pseudohemicellulose, pseudo-cellulose, and others) mainly depends on the temperature, particle size, and reactor type. The decomposition of hemicellulose and cellulose usually occurs between 470 to 530 K and 510 to 620 K, respectively, while others like lignin degrade at 550-770 K (Wang et al., 2017). The decomposition of cellulosic materials is mainly induced by depolymerization reactions forming oligosaccharides an then levoglucosan, which is finally decomposed into smaller molecules (Patwardhan et al., 2011). Similarly, the hemicellulose pyrolysis is initiated by the depolymerization of polysaccharides into oligosaccharides and, subsequently, leads to the rearrangement of the molecules, producing 1,4-Anhydro-D-xylopyranose, which further disintegrates into smaller molecular weight components.

At high temperature, secondary decomposition leads to the formation of secondary volatiles (*Volatiles2*) and biochar (*Biochar2*). This may include cracking, reforming, condensation, polymerization, oxidation, and gasification (Di Blasi, 2008). During the secondary reactions, *Biochar1* forms *Volatiles2* and *Biochar2*, as presented in Equation (13).

$$Biochar1 \xrightarrow{\frac{d\alpha_4}{dt}} Volatiles2 + Biochar2$$
(13)

Likewise, the inorganic compounds are also greatly affected during pyrolysis. These inorganic compounds, including heavy metals, may be accumulated in the cell wall and bind ionically and covalently to polysaccharides, which can initiate autocatalytic pyrolysis (Keown et al., 2005). Here, a concept of the partial gasification of *T. ornata* polluted with heavy metals from the Red Sea is presented in **Figure 2**.

These inorganic compounds may include macro/micro minerals (e.g., K, Ca, Na, Mg, Fe, Zn, and Mn) and heavy metals (e.g., Ag, Al, As, Cd, Cr, Co, Hg, Mo, Ni, Pb, and Ti). The pyrolysis process has been previously employed for processing these minerals and heavy metals for valuable products like biochar and bio-oil. For example, alkali and alkaline earth metals (such as K, Ca, and Mg) have notable catalytic effects on the pyrolytic conversion. Vaporization of K has been shown to take place at lower temperatures, while Ca and Mg vaporize at higher temperatures depending on their ionic or covalent bonding with organic molecules (Wei et al., 2012). Liu et al. (2017) reviewed biomass pyrolysis in the presence of inorganic elements including heavy metals and contributed a critical analysis for optimizing the pyrolytic mechanisms with productive resource utilization and a reduction in pyrolysis-based pollutants. Liu et al. (2017) also found that the prior literature has exhibited the enrichment of heavy metals in the solid biochar phase rather than being released into the volatile (gaseous/liquid) phase, and that there is no need to focus on the release of heavy metals in later research. Also, Ni and Cu have been previously used to enhance the quality of pyrolysis products (Richardson et al., 2013; Zhao et al., 2015). Hence, there is an urgent need to investigate the mechanisms for the use of inherited available inorganics and heavy metals to improve the processes and products (Liu et al., 2017). **Table 4** presents a few studies on biomass pyrolysis containing salts of heavy metals.

The summation of the rate equations in Equation (14) gives the overall degradation rate.

$$\frac{\mathrm{d}\alpha}{\mathrm{d}t} = \sum_{j=1}^{4} \frac{\mathrm{d}\alpha_j}{\mathrm{d}t} = \sum_{j=1}^{4} x_j \cdot A_j \cdot \exp\left(-\frac{E_{aj}}{RT}\right) \cdot f(\alpha) \qquad (14)$$

The separation of the pseudo-component peaks was carried out assuming an asymmetric Fraser–Suzuki (FS) distribution with the following function (Perejón et al., 2011):

$$\frac{\mathrm{d}\alpha_j}{\mathrm{d}t} = a_0 exp \left[-ln2 \left\{ \frac{ln \left(1 + 2a_1 \frac{T - T_m}{HW} \right)}{a_1} \right\}^2 \right]$$
(15)

where a_0 is the amplitude, a_1 is the asymmetry, T_m is the peak maximum temperature and HW is the half width of the peak.

Initial values of kinetic parameters were obtained from modelfree methods applied to the separated pseudo-components. The optimization of the kinetic parameters was performed by nonlinear least square fitting using Equation (10).

RESULTS AND DISCUSSIONS

Thermal Behavior and Reactivity Analysis

Figure 3 shows the TG and DTG profiling of T. ornata against temperature at different heating rates. T. ornata biomass was pyrolyzed to 850°C at different heating rates. The DTG curves feature two prominent zones, as illustrated in Figure 3B: A primary pyrolysis zone and a secondary reaction zone. In the primary pyrolysis zone, more than 90% of the total amount of mass loss was observed, resulting from reactions such as depolymerization, decarboxylation, and cracking. The maximum mass loss rates increased from 0.043 to 0.290%·s⁻¹ with an incremental increase in the heating rates of 5 to 30°C·min⁻¹. Secondary reaction zones (Zone II), initiated at 875 K, and the remaining mass was decomposed at a slower rate. This slow rate is caused by the trapped volatile matter present in the biochar and delays the decomposition (Richardson et al., 2013). The most common characteristic of all the DTG and TG curves is the increase in the conversion rate and the shifting of the peaks toward the higher temperatures when the heating rates are increased. This increase in conversion rates may be ascribed to the improved transport phenomenon. Meanwhile, the shifting of peaks is rendered to the slowing down of the pyrolysis process owing to the difficulty in the heat transfer from the surrounding to the sample for a shorter time and larger thermal lag (Shuping et al., 2010; Han et al., 2017; Özsin and Pütün, 2017). It can also be perceived from the DTG curves that the peaks in the secondary reaction zone (900-1,100 K) may be caused by the catalytic effect and gasification of the biochar due to the presence of inorganic oxides.



Isoconversional Methods

The pyrolytic characteristics were measured using the two isoconversional methods Friedman and KAS. **Figure 4** establishes the Arrhenius plot using Equations (5, 6) at different conversions. The data were fitted using linear regression lines for both differential and integral isoconversional methods at the progressive increase of conversion from 10 to 90% with a step size of 5%, as depicted in **Figure 4**.

The apparent activation energy E_a was measured from the slope of the straight lines from the plots. The quality of the regression models was estimated by the coefficient of determination (R^2), which strengthens our confidence in the estimation of E_a from the observed data. The mean apparent activation energy E_{α} , calculated from the Friedman and KAS isoconversional models, are 164.03 and 146.41 kJ·mol⁻¹, respectively.

Model-Fitting Methods

Model-fitting methods have two stages, namely (a) the estimation of the initial parameters and (b) the application of iterative routines to obtain the optimized values of the kinetic parameters.

Combined Kinetics

The average activation energies (E_a) obtained from the isoconversional methods (both differential and integral) were used to obtain the order-based and SB model parameters by optimizing Equations (7, 9) using MS Excel Solver. The differential best-fit curves of *T. ornata* with optimized order-based and SB model parameters at different heating rates (5, 10, 15, and 30° C·min⁻¹) are shown in **Figure 5**. It is worth noting here that this displays the fitting of the data to order-based and SB models through a single reaction. Moreover, a single heating rate

may render inaccuracies; therefore, the mean values calculated at different heating rates are presented in **Table 5**.

The SB model is a powerful tool that allows the measurement of complex degradation mechanisms (Gámiz-González et al., 2017). As is evident from Table 5 and has been previously found as well, one of the SB model parameters is numerically absent (Šesták and Berggren, 1971; Ali et al., 2018). In the present study for all heating rates, parameter *m*, which corresponds to the deceleration mechanism, was not involved and hence the pyrolysis of macro-algae (T. ornata) was most probably driven by random nucleation and growth (Wang et al., 2006; Li et al., 2010; Ye et al., 2010). Table 5 shows a Fit% higher for the SB model, which indicates that the SB model can better explain thermal degradation than the order-based model. Thus, the parameters obtained from the SB model can better represent the pyrolysis mechanism. The average values of A obtained from the Friedman method were 6.4 \times 10¹³ and 1.3 \times 10¹⁴, while for the KAS method these were 1.0×10^{12} and 2.3×10^{12} for the order-based and SB models, respectively. The values of *n* for the order-based and SB models were 7.47 and 8.08 for the Friedman method and 6.76 and 7.44 for the KAS method, respectively. The high value of *n* in the SB model with a higher Fit% (>92) highlights the roles of order-based, diffusion and geometrical contraction in T. ornata pyrolysis. The order-based mechanism can be attributed to the nucleation process in a solid state reaction where the nucleus growth is induced by nuclei collisions with each other (Poletto et al., 2012). The optimum p-values (0.40 and 0.43) signify that the degradation process is partially controlled by the random nucleation mechanism, which randomly promotes the generation of activation centers. It is highly likely that the presence of oxygen carriers (gases and inorganic salts containing heavy metals) could initiate partial gasification (Huang et al., 2013) and catalyze T. ornata pyrolysis. This would increase the

TABLE 4 | Influence of heavy metals present in biomass during pyrolysis.

Heavy metal polluted biomass	Heavy metal	Pyrolysis performance	Major findings	References
Fir sawdust	Cu	 Increase in bio-oil yields Decreased in the oxygen contents in bio-oil Significant increase in HHV Cu-FSD derived bio-oil have higher HHV of 13.42–14.79 MJ kg⁻¹ as compared to HHV of FSD which range between 10.77 and 12.32 MJ kg⁻¹ Increase in biochar yields Lower gas yields 	The increase in yields of bio-oil and biochar attributes toward catalytic effects of Cu. The decrease in gas yields may be an indication of 1) More gas composition reactions like H ₂ with lignin in the presence of Cu to form phenols 2) Cu may promote the production of aromatics to increase bio-oil yield than gas yield. A significant quantity of C7–C10 was formed because of the catalytic effect of Cu on breaking down lignin-chains.	Liu et al., 2012
Typha angustifolia	Pb	 More than 98% Pb was distributed in biochar phase while only traces of Pb were present in bio-oil. Decrease in biochar yield from 46.5 to 33.0 % whereas the gas yield increase with increase in temperature from 673 to 873 K 	The quick decline in biochar yield with the elevation of temperature may be due to cracking of the volatiles, which shows improved decomposition of biomass. An economic analysis renders leaching methods to be expensive compared to pyrolysis. This shows bio-oil produced can be used for various applications without secondary pollution.	Liu et al., 2011
Wood	Ni	 At below 500°C, the <i>in situ</i> Ni nanoparticles (Ni°NPs) were produced during pyrolysis which offers active sites to enhance the catalytic effect on biomass pyrolysis by adsorbing the aromatics. Biochar supported NiNPs was produced, which was later used in secondary reactions as a catalyst to improve tar conversions and hydrogen production. 	The catalytic effect for both <i>in-situ</i> Ni°NPs and synthesized Ni°NPs on secondary reactions were studied. Results of <i>in-situ</i> Ni°NPs pyrolysis, synthesized Ni°NPs, and catalyst-free wood pyrolysis was evaluated. The <i>in situ</i> Ni°NPs catalyst featured high production of aromatics than synthesized one. There was a clear difference in the performance of the pyrolysis when using the <i>in-situ</i> metal species present in the wood biomass because of their catalytic effect	Richardson et al., 2013
Arundo donax and Broussonetia papyrifera	Cd, Cu, Pb	 Activation Energy (<i>E_a</i>) was calculated for both the biomasses having heavy metal salts. The <i>E_a</i> tends to decrease when pyrolysis was performed with these polluted biomasses. This may be because of the catalytic effect of these heavy metals. Increase in bio-oil yields. The volatile metals increased when having these meals salts, from 55 13 to 59.81% for <i>Annoba donax</i> and from 	Cd, Cu, and Pd promoted the pyrolysis process by decreasing the activation energies of the biomasses. These metals could assist in producing more volatiles both in liquid and gas phase	Han et al., 2017



weight loss rate and increase the gases produced (like CH_4 , H_2 , and CO_2) from the pyrolysis of *T. ornata* and could remove lattice oxygen from the oxygen carriers. Depending upon the amount of inorganics accumulated by the marine algae *T. ornata*, the catalytic characteristics of heavy metals stimulating pyrolysis and partial gasification could vary.

44.62 to 52.12 % for Broussonetia papyrifera.

Independent-Parallel Reactions

From the DTG curves of samples in **Figure 3B**, it is indicated that the degradation process is taking place in a multiple reaction scheme since the peaks are overlapping in the close temperature range. The thermal decomposition of hemicellulose in *T. ornata* biomass started at about 473 K, followed by a





major mass loss between 550 and 660 K and tailing up to 900 K; then there was a slight shoulder followed by a tail in the temperature range 880–1100 K, which could have induced the partial gasification mechanism using inherited minerals. Most of the mass loss occurs in Zone (I), as shown in **Figure 3B**, which takes the shape of a shoulder with overlapping peaks, which corresponds to hemicellulose and cellulosic material, followed by the contribution of other components with an interplay of the inherited minerals present in *T. ornata* collected from the Red Sea. The shoulder and tailing in Zone (II) may correspond to the decomposition of inorganics and could proceed further to a partial gasification process. Hence, it is necessary to apply three parallel-reaction models for the primary reactions in Zone (I) pertaining to the decomposition of pseudo-hemicellulose,

pseudo-cellulose and pseudo-others. Overall, the DTG curves at 5, 10, 15, and 30° C·min⁻¹ were decomposed into three primary and one secondary decomposition reactions by employing asymmetric Fraser-Suzuki (FS) distributions, as presented in **Figure 6**. Recent studies show have shown that the application of the FS function suits the thermal degradation of a solid state reaction into its individual processes (Millán et al., 2017). The mass loss contributions (x_j) of separated peaks obtained from **Figure 6** are presented in **Table 6**.

The average relative mass loss contribution (x_j) for pseudohemicellulose, pseudo-cellulose, pseudo-others, and biochar at 5, 10, 15, and 30° C·min⁻¹ was 40.5, 13.0, 38.5, and 8.0%, respectively. It can also be seen from **Figure 5** that the degradation rates of pseudo-components increase with

β [° C	·min ⁻¹]		Order					SE	3		
		n [-]	А [s ⁻¹]	SSE [s ⁻²]	Fit [%]	А [s ⁻¹]	т [-]	n [-]	р [-]	SSE [s ⁻²]	Fit [%]
Differential	5	7.03	4.6×10^{13}	1.6×10^{-5}	90.46	1.2×10^{14}	0	7.72	0.55	1.0×10^{-5}	92.60
	10	7.43	6.7×10^{13}	2.3×10^{-5}	92.47	1.3×10^{14}	0	8.00	0.36	1.7×10^{-5}	93.50
	15	7.61	6.8×10^{13}	3.2×10^{-5}	92.46	1.2×10^{14}	0	8.13	0.31	2.6×10^{-5}	93.29
	30	7.80	7.7×10^{13}	8.6×10^{-5}	91.28	1.6×10^{14}	0	8.34	0.36	6.4×10^{-5}	92.62
	Average (± SE)	7.47 (± 0.16)	6.4×10^{13} (± 0.7 × 10 ¹³)	-	-	1.3×10^{14} (± 0.1 × 10 ¹⁴)	0	8.08 (± 0.13)	0.40 (± 0.05)	-	-
Integral	5	6.35	6.7 × 10 ¹¹	1.6×10^{-5}	90.35	1.7 × 10 ¹²	0	7.07	0.55	9.1×10^{-6}	92.93
	10	6.73	1.0×10^{12}	2.3×10^{-5}	92.19	2.2×10^{12}	0	7.38	0.41	1.6×10^{-5}	93.65
	15	6.89	1.1×10^{12}	3.3×10^{-5}	92.16	2.1×10^{12}	0	7.36	0.36	2.5×10^{-5}	93.41
	30	7.06	1.3×10^{12}	9.0×10^{-5}	90.88	1.6×10^{14}	0	7.67	0.40	6.2×10^{-5}	92.70
	Average (± SE)	6.76 (± 0.15)	1.0×10^{12} (± 0.1 × 10 ¹²)	-	-	2.3×10^{12} (± 0.3 × 10 ¹²)	0	7.44 (± 0.12)	0.43 (± 0.04)	-	-

TABLE 5 | Optimized order-based and SB model parameters at different heating rates with isoconversional methods.



increasing heating rates. Furthermore, with increasing heating rates the Fit% increases.

The major mass loss comes from the degradation of pseudohemicellulose and others, accounting for relative mass loss contributions of 40.5 and 38.5%, respectively. The decomposition continues over wide temperature ranges. The secondary decomposition in Zone (II) contributes to an additional mass loss of 8%, which results in the evolution of more volatiles. Therefore, the remaining mass after secondary decomposition, termed as *Biochar2*, further decreases. The decompositions of carbohydrates, others, and *Biochar1* in the presence of inorganics indicates the cumulative effects of catalysts and gasification. This could encourage flash pyrolysis or rapid heating in order to fully utilize the gasification and catalytic effects of these inorganically rich marine biomasses.

The kinetic parameters of the pseudo-component degradations were estimated using the Friedman and KAS model-free methods on separated deconvoluted peaks. The E_a of the independent-parallel reactions were calculated from the regression plots of Friedman and KAS and the evolution of

notional mass loss (v.) of	deconvoluted Turbinaria ornata	neoudo componente using ES	distribution at different heating rates
201101 Iai 111ass 10ss (Aj) 01	deconvoluted fundinaria offiata	pseudo-components using r o	uistribution at unrerent neating rates.

β [°C⋅min ^{−1}]	Pseudo- Hemicellulose (Peak 1)	Pseudo- Cellulose (Peak 2)	Others (Peak 3)	<i>Biochar1</i> (Peak 4)	SSE [s ⁻²]	Fit [%]
5	0.425	0.147	0.361	0.067	2.3×10^{-7}	98.85
10	0.403	0.124	0.397	0.075	3.1×10^{-7}	99.14
15	0.397	0.107	0.411	0.085	4.4×10^{-7}	99.16
30	0.394	0.143	0.370	0.093	8.2×10^{-7}	99.23
Average (± SE)	0.405 (+ 0.007)	0.130 (+ 0.009)	0.385 (+ 0.012)	0.080 (+ 0.006)	-	-



TABLE 7 | Optimized kinetic parameters of pseudo-components.

β [°C⋅min [−]	.1]	Pseude	o-hemicellu	lose	Pseu	ido-cellulos	e		Others			Biochar			
		A [s ⁻¹]	<i>E_a</i> [kJ⋅mol ⁻¹]	n] [-]	A [s ⁻¹]	<i>E_a</i> [kJ⋅mol ⁻¹]	n [-]	A [s ⁻¹]	<i>E_a</i> [kJ⋅mol ⁻¹]	n [-]	A [s ⁻¹]	<i>E_a</i> [kJ⋅mol ⁻¹]	n [-]	SSE [s ⁻²]	Fit [%]
Differential	5	1.9 × 10 ¹⁹	223.9	2.52	5.1 × 10 ¹⁶	224.6	1.48	0.02	23.7	1.10	2.2 × 10 ¹³	309.7	1.79	1.3×10^{-6}	97.50
	10	1.4×10^{12}	151.8	1.79	3.4×10^{16}	223.1	1.28	0.04	25.4	0.97	1.6×10^{13}	304.9	2.44	6.0×10^{-7}	98.83
	15	3.9×10^{14}	176.1	2.10	1.8×10^{13}	186.2	1.12	0.08	27.1	1.09	9.6×10^5	158.6	1.24	1.1×10^{-6}	98.71
	30	2.2×10^{15}	182.8	1.97	2.5×10^{13}	186.8	1.14	0.12	27.0	0.96	5.0×10^5	187.3	1.31	2.1×10^{-6}	98.80
Integral	5	0.9 × 10 ¹⁹	230.6	2.62	2.1 × 10 ¹⁵	209.4	1.38	0.02	24.6	1.18	2.7 × 10 ¹⁰	258.9	1.54	1.3×10^{-6}	97.43
	10	1.5×10^{20}	231.8	2.71	2.3×10^{15}	209.6	1.34	0.02	21.3	1.09	1.5×10^{10}	252.8	1.79	3.1×10^{-6}	97.49
	15	1.0×10^{13}	160.3	1.88	1.4×10^{15}	207.6	1.19	0.05	25.1	0.94	2.4×10^{10}	253.9	1.98	9.7×10^{-7}	98.78
	30	2.8×10^{16}	194.0	2.09	2.6×10^{13}	187.0	1.17	0.20	29.3	1.08	2.6×10^6	181.5	1.33	2.4×10^{-6}	98.73

activation energies with respect to conversion; these are shown in **Figure 7**.

It is clear from **Figure 7** that the primary decompositions overlap and that the interplay of the components causes the overall activation energy decrease until around 90% of the conversion. This aspect has been regarded by some researchers to the condensation of polymers at higher temperature to form char (Chorley, 1892; Basile et al., 2016; Anca-Couce and Scharler, 2017; Ciuta et al., 2018). The secondary conversion of *Biochar1* at a higher temperature with a higher activation energy, which drops as the conversion proceeds toward completion. Average activation energies of the pseudo-components and initial parameter values obtained from the Friedman and KAS methods were used to optimize the kinetic parameters via non-linear

regression. The SB parameters m and p were absent, hence both order-based and SB models were found to be equivalent. The obtained parameters using differential and integral methods at different heating rates are presented in **Table 7**.

A summary of the key thermal and kinetic parameters is tabulated in **Table 8**. It is evident that the peak-maximum temperature (T_m) for the decomposition of pseudo-hemicellulose is 541.2 K for the differential Friedman method and 541.5 K for the integral KAS method. The T_m for pseudo-cellulose is 598.4 and 583.4 K using the Friedman and KAS methods, respectively. Pseudo-others decompose with a T_m of 641.8 to 634.2 K according to the Friedman and KAS methods. The decomposition at mild-to-high temperatures may also involve condensation of polymers. Depending on the heating rate, the

Components	Fried (Di	man meth ifferential)	od				КА (S method Integral)	l			
	τ _m	αm	× _j	n	A	<i>E_a</i>	<i>T_m</i>	αm	× _j	n	A	<i>Ea</i>
	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)	(± SE)
	[K]	[%]	[−]	[-]	[s ⁻¹]	[kJ.mol ⁻¹] [K]	[%]	[−]	[-]	[s ⁻¹]	[kJ.mol ⁻¹]
Pseudo-	541.2	26.6	0.351	2.09	4.7×10^{18}	183.7	541.5	26.6	0.333	2.33	5.9×10^{19}	204.2
hemicellulose	(± 3.6)	(± 0.7)	(± 0.009)	(± 0.16)	(± 4.7 × 10 ¹⁸)	(± 14.97)	(± 3.6)	(± 0.8)	(± 0.010)	(± 0.20)	(± 3.6 × 10 ¹⁹)	(± 17.05)
Pseudo-	598.4	61.3	0.164	1.03	2.1 × 10 ¹⁶	205.2	583.4	61.6	0.162	1.27	1.5×10^{15}	203.4
cellulose	(± 5.3)	(± 0.0)	(± 0.004)	(± 0.07)	(± 1.3 × 10 ¹⁶)	(± 10.8)	(± 14.2)	(± 0.0)	(± 0.006)	(± 0.05)	(± 0.5 × 10 ¹⁵)	(± 5.5)
Others	641.8	70.9	0.408	1.03	0.06	25.8	634.2	71.8	0.429	1.07	0.07	25.1
	(± 13.7)	(± 0.0)	(± 0.008)	(± 0.04)	(± 0.02)	(± 0.8)	(± 15.3)	(± 0.0)	(± 0.016)	(± 0.05)	(± 0.04)	(± 1.6)
Biochar1	963.9	96.0	0.077	1.69	9.5×10^{12}	240.1	964.2	96.1	0.076	1.66	1.6×10^{10}	236.8
	(± 10.9)	(± 0.0)	(± 0.008)	(± 0.28)	(± 5.6 × 10 ¹²)	(± 39.2)	(± 10.0)	(± 0.0)	(± 0.007)	(± 0.14)	(± 0.6 × 10 ¹⁰)	(± 18.5)
Weighted Average		-	-	1.45	1.7 × 10 ¹⁸	127.15		-	-	1.57	2.0 × 10 ¹⁹	129.70

secondary decomposition of *Biochar1* occurs in a temperature range of 870-1,100 K with a T_m of 963.9 and 964.2 K using the Friedman and KAS methods, respectively.

The average reaction order (*n*) of the decomposition of pseudo-hemicellulose, pseudo-cellulose, others, and biochar was found to be 2.09, 1.03, 1.03, and 1.69 for the Friedman method and 2.33, 1.27, 1.07, and 1.66 for the KAS method, respectively. Similarly, the average pre-exponential factor (*A*) of the decomposition of pseudo-hemicellulose, pseudo-cellulose, others, and biochar was found to be 4.7×10^{18} , 2.1×10^{16} , 0.06, and $9.5 \times 10^{12} \text{ s}^{-1}$, respectively, for the Friedman method and 5.9×10^{19} , 1.5×10^{15} , 0.07, and 1.6×10^{10} for the KAS method, respectively.

The average E_a of pseudo-hemicellulose obtained from the Friedman and KAS methods was 183.7 and 204.2 kJ.mol⁻¹, respectively. Similarly, the average E_a of pseudo-cellulose using the Friedman and KAS methods was 205.2 and 203.4 kJ.mol⁻¹, respectively. Pseudo-others decomposed with a lower average E_a of 25.8 and 25.1 kJ.mol⁻¹, as estimated from the Friedman and KAS methods, respectively. The average E_a of pseudo-others is very low, which may be due to the formation of polymers owing to the metal oxides from inherited inorganics taken up by T. ornata macro-algae. Jian et al. (2014) found that the E_a of metal oxides was low at higher heating rates, indicating oxygen transport to be a rate limiting step at high temperatures and high heating rates. This may induce a partial gasification phenomenon at this stage. The partial gasification, according to Equation (8), could be due to the inherited metal oxides present in the cell wall of the polysaccharide matrix of T. ornata. The average E_a was calculated for the degradation of *Biochar1*, which was 240.1 and 236.5 kJ.mol⁻¹ using the Friedman and KAS methods, respectively. It can also be deduced that the

REFERENCES

Abdallah, A. M. A., Abdallah, M. A., Beltagy, A., and Siam, E. (2006). Contents of heavy metals in marine algae from Egyptian Red Sea coast. *Toxicol. Environ. Chem.* 88, 9–22. doi: 10.1080/02772240500414911 variation in the values of A, n, and E_a was the greatest during the secondary decomposition of *Biochar1* in Zone (II) among the other pseudo-components. Moreover, the sharp decrease in the E_a of *Biochar1* is a mere indication of the spontaneity of the process.

CONCLUSIONS

Marine macro-algae contains inorganics, including heavy metals, the amount of which depends on the macro-algae's ability to uptake these as well as the concentration of inorganics dissolved in the marine environment. While this help mitigates the heavy metals, their disposal needs care. Pyrolysis offers not only a means to handle such materials but also the ability to use these to our benefit in the form of valuable products such as volatiles and biochar. The Red Sea is a rich source of many minerals, which are taken up by the organisms living there. The role of these organisms in this process is important and needs to be evaluated using complex models. A four-component independent-reaction model using a simple order-based mechanism for each reaction is sufficient to analyze the pyrolytic kinetics. This shows that three primary reactions proceed simultaneously, producing a solid product which degrades through a secondary decomposition reaction at a higher temperature. Interestingly, the activation energy drops considerably for the secondary reaction, hinting at the partial gasification and the catalytic effects of the minerals inherent in the macro-algae.

AUTHOR CONTRIBUTIONS

All authors listed have made a substantial, direct and intellectual contribution to the work, and approved it for publication.

Aboyade, A. O., Carrier, M., Meyer, E. L., Knoetze, J. H., and Görgens, J. F. (2012). Model fitting kinetic analysis and characterisation of the devolatilization of coal blends with corn and sugarcane residues. *Thermochim. Acta* 530, 95–106. doi: 10.1016/j.tca.2011. 12.007

- Adanez, J., Abad, A., Garcia-Labiano, F., Gayan, P., and de Diego, L. F. (2012). Progress in chemical-looping combustion and reforming technologies. *Prog. Energy Combust. Sci.* 38, 215–282. doi: 10.1016/j.pecs.2011.09.001
- Agency, I. E. (2017). Energy Technology Perspectives 2017: Catalysing Energy Technology Transformations. OECD/IEA.
- Ali, I., Naqvi, S. R., and Bahadar, A. (2018). Kinetic analysis of *Botryococcus braunii* pyrolysis using model-free and model fitting methods. *Fuel* 214, 369–380. doi: 10.1016/j.fuel.2017.11.046
- Anca-Couce, A., and Scharler, R. (2017). Modelling heat of reaction in biomass pyrolysis with detailed reaction schemes. *Fuel* 206, 572–579. doi: 10.1016/j.fuel.2017.06.011
- Baghel, R. S., Mantri, V. A., and Reddy, C., R. K. (2017). "A new wave of research interest in marine macroalgae for chemicals and fuels: challenges and potentials," in *Fuels, Chemicals and Materials from the Oceans and Aquatic Sources*, eds. F. M. Kerton, and N. Yan (John Wiley and Sons, Ltd), 43–63. doi: 10.1002/9781119117193.ch3
- Basile, L., Tugnoli, A., Stramigioli, C., and Cozzani, V. (2016). Thermal effects during biomass pyrolysis. *Thermochim. Acta* 636, 63–70. doi: 10.1016/j.tca.2016.05.002
- Behairy, A. K. A., and El-Sayed, M. M. (1983). Biochemical Composition of Some Marine Brown Algae From Jeddah Coast, Saudi Arabia. IJMS Vol.12(3) [September 1983]. Available online at: http://nopr.niscair.res.in/ handle/123456789/38893 (Accessed December 17, 2017).
- Bold, H. C., and Wynne, M. J. (1984). *Introduction to the Algae, 2nd Edn.* Englewood Cliffs, NJ: Prentice-Hall.
- Bruckner, A., Rowlands, G., Riegl, B., Purkis, S., Williams, A., and Renaud, P. (2013). Atlas of Saudi Arabian Red Sea Marine Habitats. Oceanography Faculty Books and Book Chapters. Available online at: http://nsuworks.nova.edu/occ_ facbooks/33
- Chorley, C. (1892). The destructive distillation of wood. J. Soc. Chem. Industry 11, 872–874. doi: 10.1002/jctb.5000111104
- Ciuta, S., Tsiamis, D., and Castaldi, M. J. (2018). "Chapter two fundamentals of gasification and pyrolysis," in *Gasification of Waste Materials*, eds S. Ciuta, D. Tsiamis, and M. J. Castaldi (Academic Press), 13–36. doi: 10.1016/B978-0-12-812716-2.00002-9
- Davis, T. A., Volesky, B., and Mucci, A. (2003). A review of the biochemistry of heavy metal biosorption by brown algae. *Water Res.* 37, 4311–4330. doi: 10.1016/S0043-1354(03)00293-8
- Deepak, P., Sowmiya, R., Balasubramani, G., and Perumal, P. (2017a). Phytochemical profiling of Turbinaria ornata and its antioxidant and anti-proliferative effects. J. Taibah Univ. Med. Sci. 12, 329–337. doi: 10.1016/j.jtumed.2017.02.002
- Deepak, P., Sowmiya, R., Ramkumar, R., Balasubramani, G., Aiswarya, D., and Perumal, P. (2017b). Structural characterization and evaluation of mosquitolarvicidal property of silver nanoparticles synthesized from the seaweed, Turbinaria ornata (Turner) J. Agardh 1848. Artif. Cells Nanomed. Biotechnol. 45, 990–998. doi: 10.1080/21691401.2016.1198365
- Di Blasi, C. (2008). Modeling chemical and physical processes of wood and biomass pyrolysis. *Progr. Energy Combust. Sci.* 34, 47–90. doi: 10.1016/j.pecs.2006.12.001
- Dussan, K., Dooley, S., and Monaghan, R. (2017). Integrating compositional features in model compounds for a kinetic mechanism of hemicellulose pyrolysis. *Chem. Eng. J.* 328, 943–961. doi: 10.1016/j.cej.2017.07.089
- El-Naggar, M. E. E., and Al-Amoudi, O. A. (1989). Heavy metal levels in several species of marine algae from the Red Sea of Saudia Aabia. J. King Abdulaziz Univ. Sci. J. 1, 5–13. doi: 10.4197/Sci.1-1.1
- Ferdouse, F., Holdt, S. L., Smith, R., Murúa, P., and Yang, Z. (2018). *The Global Status of Seaweed Production, Trade and Utilization*. Rome, Italy: Products, Trade and Marketing Branch, Fisheries and Aquaculture Policy and Resources Division, Food and Agriculture Organization of the United Nations Available online at: http://www.fao.org/in-action/globefish/ publications/details-publication/en/c/1154074/ (Accessed Jan 26, 2019).
- Friedman, H. L. (1964). Kinetics of thermal degradation of char-forming plastics from thermogravimetry. application to a phenolic plastic. J. Polym. Sci. C Polym. Symp. 6, 183–195.
- Gámiz-González, M. A., Correia, D. M., Lanceros-Mendez, S., Sencadas, V., Gómez Ribelles, J. L., and Vidaurre, A. (2017). Kinetic study of thermal degradation of chitosan as a function of deacetylation degree. *Carbohydr. Polym.* 167, 52–58. doi: 10.1016/j.carbpol.2017.03.020

- Ghadiryanfar, M., Rosentrater, K. A., Keyhani, A., and Omid, M. (2016). A review of macroalgae production, with potential applications in biofuels and bioenergy. *Renew. Sustain. Energy Rev.* 54, 473–481. doi: 10.1016/j.rser.2015.10.022
- Grigiante, M., Brighenti, M., and Antolini, D. (2016). A generalized activation energy equation for torrefaction of hardwood biomasses based on isoconversional methods. *Renew. Energy* 99, 1318–1326. doi: 10.1016/j.renene.2016.07.054
- Han, Z., Guo, Z., Zhang, Y., Xiao, X., Xu, Z., and Sun, Y. (2017). Pyrolysis characteristics of biomass impregnated with cadmium, copper and lead: influence and distribution. *Waste Biomass Valor* 2017, 1–8. doi: 10.1007/s12649-017-0036-5
- Hong, Y., Chen, W., Luo, X., Pang, C., Lester, E., and Wu, T. (2017). Microwaveenhanced pyrolysis of macroalgae and microalgae for syngas production. *Bioresour. Technol.* 237, 47–56. doi: 10.1016/j.biortech.2017.02.006
- Huang, Z., Deng, Z., He, F., Chen, D., Wei, G., Zhao, K., et al. (2017). Reactivity investigation on chemical looping gasification of biomass char using nickel ferrite oxygen carrier. *Int. J. Hydrogen Energy* 42, 14458–14470. doi: 10.1016/j.ijhydene.2017.04.246
- Huang, Z., He, F., Feng, Y., Liu, R., Zhao, K., Zheng, A., et al. (2013). Characteristics of biomass gasification using chemical looping with iron ore as an oxygen carrier. *Int. J. Hydrogen Energy* 38, 14568–14575. doi: 10.1016/j.ijhydene.2013.09.022
- Jian, G., Zhou, L., Piekiel, N. W., and Zachariah, M. R. (2014). Low effective activation energies for oxygen release from metal oxides: evidence for mass-transfer limits at high heating rates. *ChemPhysChem* 15, 1666–1672. doi: 10.1002/cphc.201301148
- Kayalvizhi, K., Asmathunisha, N., Subramanian, V., and Kathiresan, K. (2014). Purification of silver and gold nanoparticles from two species of brown seaweeds (*Padina tetrastromatica* and *Turbinaria ornata*). J. Med. Plants Stud. 2, 32–37.
- Keown, D. M., Favas, G., Hayashi, J., and Li, C. Z. (2005). Volatilisation of alkali and alkaline earth metallic species during the pyrolysis of biomass: differences between sugar cane bagasse and cane trash. *Bioresour. Technol.* 96, 1570–1577. doi: 10.1016/j.biortech.2004.12.014
- Kissinger, H. E. (1957). Reaction kinetics in differential thermal analysis. Anal. Chem. 29, 1702–1706 doi: 10.1021/ac60131a045
- Li, D., Chen, L., Yi, X., Zhang, X., and Ye, N. (2010). Pyrolytic characteristics and kinetics of two brown algae and sodium alginate. *Bioresour. Technol.* 101, 7131–7136. doi: 10.1016/j.biortech.2010.03.145
- Liu, W. J., Li, W. W., Jiang, H., and Yu, H. Q. (2017). Fates of chemical elements in biomass during its pyrolysis. *Chem. Rev.* 117, 6367–6398. doi: 10.1021/acs.chemrev.6b00647
- Liu, W. J., Tian, K., Jiang, H., Zhang, X. S., Ding, H. S., and Yu, H. Q. (2012). Selectively improving the bio-oil quality by catalytic fast pyrolysis of heavymetal-polluted biomass: take copper (Cu) as an example. *Environ. Sci. Technol.* 46, 7849–7856. doi: 10.1021/es204681y
- Liu, W. J., Zeng, F. X., Jiang, H., Zhang, X. S., and Yu, H. Q. (2011). Technoeconomic evaluation of the integrated biosorption–pyrolysis technology for lead (Pb) recovery from aqueous solution. *Bioresour. Technol.* 102, 6260–6265. doi: 10.1016/j.biortech.2011.02.104
- Maceiras, R., Rodri'guez, M., Cancela, A., Urréjola, S., and Sánchez, A. (2011). Macroalgae: raw material for biodiesel production. *Appl. Energy* 88, 3318–3323. doi: 10.1016/j.apenergy.2010.11.027
- Manivannan, K., Thirumaran, G., Devi, G. K., Anantharaman, P., and Balasubramanian, T. (2009). Proximate composition of different group of seaweeds from Vedalai coastal waters (Gulf of Mannar): southeast coast of India. *Middle East J. Sci. Res.* 4, 72–77.
- Millán, L. M. R., Vargas, F. E. S., and Nzihou, A. (2017). Kinetic analysis of tropical lignocellulosic agrowaste pyrolysis. *Bioenerg. Res.* 10, 832–845. doi: 10.1007/s12155-017-9844-5
- Mohamed Ali, D., and Elnabawy Ward, F. M. (2016). Ecological and biochemical analyses of the brown alga Turbinaria ornata (Turner) J. Agardh from Red Sea coast, Egypt. J. Coastal Life Med. 4, 187–192. doi: 10.12980/jclm.4.2016j6-10
- Mohammadi, M., Tajik, H., and Hajeb, P. (2013). Nutritional composition of seaweeds from the Northern Persian Gulf. *Iran. J. Fish. Sci.* 12, 232–240.
- Mohan, D., Pittman Charles, U., and Steele, P. H. (2006). Pyrolysis of wood/biomass for bio-oil: a critical review. *Energy Fuels* 20, 848–889. doi: 10.1021/ef0502397

- Oliveira, J. (2016). An Integrated Use of Macroalgae as Bioproducts Source and Biosorbent for Environmental Applications. Porto: Faculdade de Engenharia da Universidade do Porto.
- Omar, H. H., Shiekh, H. M., Gumgumjee, N. M., El-Kazan, M. M., and El-Gendy, A. M. (2012). Antibacterial activity of extracts of marine algae from the Red Sea of Jeddah, Saudi Arabia. *Afr. J. Biotechnol.* 11, 13576–13585. doi: 10.5897/AJB12.780
- Opfermann, J. R., Kaisersberger, E., and Flammersheim, H. J. (2002). Model-free analysis of thermoanalytical data-advantages and limitations. *Thermochimica Acta* 391, 119–127. doi: 10.1016/S0040-6031(02)00169-7
- Orfão, J. J. M., Antunes, F. J. A., and Figueiredo, J. L. (1999). Pyrolysis kinetics of lignocellulosic materials—three independent reactions model. *Fuel* 78, 349–358 doi: 10.1016/S0016-2361(98)00156-2
- Ortiz-Calderon, C., Silva, H. C., and Vásquez, D. B. (2017). "Metal removal by seaweed," in *Biomass Volume Estimation and Valorization for Energy*, ed J. S. Tumuluru (InTechOpen), 361–380.
- Osman, A. I., Abdelkader, A., Johnston, C. R., Morgan, K., and Rooney, D. W. (2017). Thermal investigation and kinetic modeling of lignocellulosic biomass combustion for energy production and other applications. *Ind. Eng. Chem. Res.* 56, 12119–12130. doi: 10.1021/acs.iecr.7b03478
- Özsin, G., and Pütün, A. E. (2017). Insights into pyrolysis and copyrolysis of biomass and polystyrene: thermochemical behaviors, kinetics and evolved gas analysis. *Energy Convers. Manage.* 149, 675–685. doi: 10.1016/j.enconman.2017.07.059
- Patwardhan, P. R., Dalluge, D. L., Shanks, B. H., and Brown, R. C. (2011). Distinguishing primary and secondary reactions of cellulose pyrolysis. *Bioresour. Technol.* 102, 5265–5269. doi: 10.1016/j.biortech.2011.02.018
- Perejón, A., Sánchez-Jiménez, P. E., Criado, J. M., and Pérez-Maqueda, L. A. (2011). Kinetic analysis of complex solid-state reactions. A new deconvolution procedure. J. Phys. Chem. B 115, 1780–1791. doi: 10.1021/jp110895z
- Poletto, M., Zattera, A. J., and Santana, R. M. (2012). Thermal decomposition of wood: Kinetics and degradation mechanisms. *Bioresour. Technol.* 126, 7–12. doi: 10.1016/j.biortech.2012.08.133
- Ponnan, A., Ramu, K., Marudhamuthu, M., Marimuthu, R., Siva, K., and Kadarkarai, M. (2017). Antibacterial, antioxidant and anticancer properties of Turbinaria conoides (J. Agardh) Kuetz. Clin. Phytosci. 3:5. doi: 10.1186/s40816-017-0042-y
- Rajeshkumar, S., Malarkodi, C., Gnanajobitha, G., Paulkumar, K., Vanaja, M., Kannan, C., et al. (2013). Seaweed-mediated synthesis of gold nanoparticles using Turbinaria conoides and its characterization. *J. Nanostruct. Chem.* 3:44. doi: 10.1186/2193-8865-3-44
- Ranzi, E., Debiagi, P. E. A., and Frassoldati, A. (2017). Mathematical modeling of fast biomass pyrolysis and bio-oil formation. note I: kinetic mechanism of biomass pyrolysis. ACS Sustain. Chem. Eng. 5, 2867–2881. doi: 10.1021/acssuschemeng.6b03096
- Richardson, Y., Motuzas, J., Julbe, A., Volle, G., and Blin, J. (2013). Catalytic investigation of *in situ* generated Ni metal nanoparticles for tar conversion during biomass pyrolysis. *J. Phys. Chem. C* 117, 23812–23831. doi: 10.1021/jp408191p
- Roesijadi, G., Jones, S., Snowden-Swan, L., and Zhu, Y. (2010). Macroalgae as a Biomass Feedstock: A Preliminary Analysis. U.S. Department of Energy. doi: 10.2172/1006310
- Ross, A. B., Jones, J. M., Kubacki, M. L., and Bridgeman, T. (2008). Classification of macroalgae as fuel and its thermochemical behaviour. *Bioresour. Technol.* 99, 6494–6504. doi: 10.1016/j.biortech.2007.11.036
- Roy, S., P, A., Attia, A. M., and Attia, A. A. A. (2017). Heavy Metals Accumulation of Different Parts of *Turbinaria* spp. along the Olaikuda Coast, Rameshwaram, Tamilnadu, India. *IARJSET* 4, 99–102. doi: 10.17148/IARJSET.2017.4319
- Saha, B., and Orvig, C. (2010). Biosorbents for hexavalent chromium elimination from industrial and municipal effluents. *Coordinat. Chem. Rev.* 254, 2959–2972. doi: 10.1016/j.ccr.2010.06.005
- Santoso, J., Gunji, S., Yoshie-Stark, Y., and Suzuki, T. (2006). Mineral contents of indonesian seaweeds and mineral solubility affected by basic cooking. *Food Sci. Technol. Res.* 12, 59–66. doi: 10.3136/fstr.12.59
- Šesták, J., and Berggren, G. (1971). Study of the kinetics of the mechanism of solid-state reactions at increasing temperatures. *Thermochim. Acta* 3, 1–12 doi: 10.1016/0040-6031(71)85051-7

- Sheldon, R. A. (2014). Green and sustainable manufacture of chemicals from biomass: state of the art. Green Chem. 16, 950–963. doi: 10.1039/C3GC41935E
- Shuping, Z., Yulong, W., Mingde, Y., Chun, L., and Junmao, T. (2010). Pyrolysis characteristics and kinetics of the marine microalgae *Dunaliella tertiolecta* using thermogravimetric analyzer. *Bioresour. Technol.* 101, 359–365. doi: 10.1016/j.biortech.2009.08.020
- Siddall, M., Smeed, D. A., Hemleben, C., Rohling, E. J., Schmelzer, I., and Peltier, W. R. (2004). Understanding the Red Sea response to sea level. *Earth Planet. Sci. Lett.* 225, 421–434. doi: 10.1016/j.epsl.2004.06.008
- Šimon, P. (2011). Fourty years of the Šesták–Berggren equation. Thermochim. Acta 520, 156–157. doi: 10.1016/j.tca.2011.03.030
- Spalding, M. D., Green, E. P., and Ravilious, C. (2001). World Atlas of Coral Reefs, 1st Edn. Berkeley, CA: University of California Press.
- Stranska-Zachariasova, M., Kurniatanty, I., Gbelcova, H., Jiru, M., Rubert, J., Nindhia, T. G., et al. (2017). Bioprospecting of turbinaria macroalgae as a potential source of health protective compounds. *Chem. Biodivers.* 14:201600192. doi: 10.1002/cbdv.201600192
- Tenorio-Rodriguez, P. A., Murillo-Álvarez, J. I., Campa-Cordova, Á. I., and Angulo, C. (2017). Antioxidant screening and phenolic content of ethanol extracts of selected Baja California Peninsula macroalgae. J. Food Sci. Technol. 54, 422–429. doi: 10.1007/s13197-016-2478-3
- Vyazovkin, S., Burnham, A. K., Criado, J. M., Pérez-Maqueda, L. A., Popescu, C., and Sbirrazzuoli, N. (2011). ICTAC Kinetics Committee recommendations for performing kinetic computations on thermal analysis data. *Thermochim. Acta* 520, 1–19. doi: 10.1016/j.tca.2011.03.034
- Vyazovkin, S., and Wight, C. A. (1998). Isothermal and non-isothermal kinetics of thermally stimulated reactions of solids. *Int. Rev. Phys. Chem.* 17, 407–433. doi: 10.1080/014423598230108
- Wang, J., Wang, G., Zhang, M., Chen, M., Li, D., Min, F., et al. (2006). A comparative study of thermolysis characteristics and kinetics of seaweeds and fir wood. *Process Biochem.* 41, 1883–1886. doi: 10.1016/j.procbio.2006. 03.018
- Wang, S., Dai, G., Yang, H., and Luo, Z. (2017). Lignocellulosic biomass pyrolysis mechanism: a state-of-the-art review. *Prog. Energy Combust. Sci.* 62, 33–86. doi: 10.1016/j.pecs.2017.05.004
- Wei, X., Zhang, G., Cai, Y., Li, L., and Li, H. (2012). The volatilization of trace elements during oxidative pyrolysis of a coal from an endemic arsenosis area in southwest Guizhou, China. J. Analyt. Appl. Pyrol. 98, 184–193. doi: 10.1016/j.jaap.2012.08.015
- Xu, T., Xu, F., Hu, Z., Chen, Z., and Xiao, B. (2017). Non-isothermal kinetics of biomass-pyrolysis-derived-tar (BPDT) thermal decomposition via thermogravimetric analysis. *Energy Convers. Manage.* 138, 452–460. doi: 10.1016/j.enconman.2017.02.013
- Ye, N., Li, D., Chen, L., Zhang, X., and Xu, D. (2010). Comparative studies of the pyrolytic and kinetic characteristics of maize straw and the seaweed ulva pertusa. *PLoS ONE* 5:e0012641. doi: 10.1371/journal.pone.0012641
- Zhao, H., Guo, L., and Zou, X. (2015). Chemical-looping auto-thermal reforming of biomass using Cu-based oxygen carrier. *Appl. Energy* 157, 408–415. doi: 10.1016/j.apenergy.2015.04.093
- Zubia, M., Payri, C. E., Deslandes, E., and Guezennec, J. (2005). Chemical Composition of Attached and Drift Specimens of Sargassum mangarevense and *Turbinaria ornata* (Phaeophyta: Fucales) from Tahiti, French Polynesia. *Bot. Mar.* 46, 562–571. doi: 10.1515/BOT.2003.059

Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

The handling Editor declared a shared affiliation at the time of review, though no other collaboration, with the authors.

Copyright © 2019 Ali and Bahadar. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.