

A mathematical model for carbon fixation and nutrient removal by an algal photobioreactor

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Abstract

A comprehensive mathematical modeling method for *Chlorella vulgaris* (*Cv*) has been developed to assess the influence of nutrient concentration (N = 28-207 and P = 6-8 mg L⁻¹) and UV irradiation intensity ($I = 100-250 \mu\text{E}$) at feed gas CO₂ concentrations ($C_{c,g}$) of 0.04-5 %. The model encompasses gas-to-liquid mass transfer, algal uptake of carbon dioxide (C_d), nutrient removal efficiency (RE for N and P with reference to total nitrogen TN and total phosphorus TP), and the growth biokinetics of *Cv* with reference to the specific growth rate μ in d⁻¹).

The model was validated using experimental data on the *Cv* species growth in an externally illuminated photobioreactor (PBR). The fitted parameters of the model were found to be in good agreement with experimental data obtained over the range of cultivation conditions explored. The mathematical model accurately reproduced the dynamic profiles of the algal biomass and nutrient (TN and TP) concentrations, and light attenuation at different input $C_{c,g}$ values. The proposed model may therefore be used for predicting algal growth and nutrient RE for this algal species, permitting both process optimization and scale-up.

Keywords:

Mathematical model, *Chlorella vulgaris*, photobioreactor, light attenuation, Growth kinetics

Symbols

<u>Symbol</u>	<u>Description and units</u>
C	Dissolved carbon concentration, mg L ⁻¹
C_s	Saturated concentration of CO ₂ , mg L ⁻¹
D_C, D_O	Diffusivity of CO ₂ , oxygen, m ² .s ⁻¹
$H_{C,O}$	Henry constant for O ₂ or CO ₂
I	Light incident, $\mu\text{E m}^{-2}\text{s}^{-1}$
K	Light extinction coefficient, g m ⁻²
K_c	Chlorophyll-base light extinction coefficient of algae, cm ² (mg Chl-a).
k_C, k_O	Mass transfer coefficient of CO ₂ , oxygen, d ⁻¹
K_d	Biomass loss (death) rate, d ⁻¹
K_i	Carbon dissociation constant of species i , mol L ⁻¹
K_N	half saturation constants for N
K_P	half saturation constants for P
K_{TC}	half saturation constant for total carbon

L_{av}	Photosynthetically active radiation (PAR), $\mu\text{E m}^{-2}\text{s}^{-1}$
K_L	PAR half saturation constants, $\mu\text{E m}^{-2}\text{s}^{-1}$
MW_i	Molecular weight of species i , g mol^{-1}
n	Shape factor
P	Pressure, bar
r	Photobioreactor radius, m
R	Universal gas constant, $\text{L}^3 \text{bar}^{-1} \text{k}^{-1} \text{mol}^{-1}$
S_i	Concentration of selected nutrient, mg L^{-1}
T	Temperature, $^{\circ}\text{K}$
V	Volume, L
X	Biomass concentration, g L^{-1}
y	Mole fraction of CO_2 in gas phase
Y_{CTOT}	Yield coefficient for total carbon, ($\text{g}_c \text{g}_x^{-1}$)
Y_i	i nutrient yield coefficient, ($\text{g}_i \text{g}_{\text{biomass}}^{-1}$)
Y_{O_2}	Oxygen yield coefficient, ($\text{g}_{\text{O}_2} \text{g}_{\text{biomass}}^{-1}$)

Greek characters

ε	Gas holdup volume, L
$\gamma_{w,i}$	Half saturation constant for i nutrient $\gamma_{w,i}$
μ_{max}	Maximum specific growth rate, d^{-1}
θ	Angle of incident light, ($^{\circ}$)

Subscripts

g	Gas phase
l	Liquid phase
tot	Total concentration
R	Reactor
$init$	Initial value
atm	Atmospheric
$feed$	Feed

1. Introduction

1.1. Algal photobioreactor key system facets

Environmental factors such as temperature, nutrient load and UV irradiation have a direct influence on biomass productivity in an algal photobioreactor (PBR). Effective and efficient microalgae cultivation relies on a number of fundamental system properties which include (i) distribution of UV dose throughout the biomass, (ii) enhanced CO_2 mass transfer from the gas to liquid phase, (iii) enhanced CO_2 assimilation by the algal biomass (largely achieved through (i)), and (iv) effective removal of O_2 generated during the photosynthetic process which may otherwise inhibit the algal growth [1]. Carbon dioxide functions as the carbon source for most algae, with the assimilated carbon contributing about 50% of the algal biomass. The local carbon dioxide concentration at any point of a bubble column PBR should be above the minimum threshold to sustain photosynthesis to avoid carbon limitation [2].

As with all microorganisms, algal species demand various nutrients to support growth, of which nitrogen (N) and phosphorus (P) are the most important. The total N and P concentrations (TN and TP), relative to that of the algal biomass, may thus determine both algal growth and the corresponding nutrient depletion rate through their bio-assimilation, as sustained by the UV light irradiation and the availability of dissolved CO_2 [3]. Nitrogen starvation conditions have been extensively applied for maintaining

metabolic fluxes to lipids [4] leading to protein synthesis, the excess photosynthesized carbon being stored as triacylglycerides or starch [5].

Light intensity and photoperiod cycle (i.e. the relative durations of the light and dark periods) are crucial factors in determining algal growth rate, especially for photoautotrophic cultures [6]. Inefficiencies arise when microalgae are exposed to light intensities above the saturation limits, as a result of photo-inhibition or overheating. Against this, at high algal cell densities commensurately higher light intensities are required to ensure light penetration through the bulk of the culture [7].

Microalgal PBR processes thus present a number of challenges for control and optimization. Mathematical models are thus needed which are capable of quantifying the impact of practical system parameters such as bioreactor configuration, CO₂ mass transfer, carbon and nutrient uptake and water quality (pH, temperature, etc) on microalgal growth.

1.2. Previous mathematical model studies

Many different studies have been undertaken to establish a mathematical model to successfully predict algal growth in batch system. These have included:

- the fitting of experimental data to a biokinetic model for a batch PBR [8];
- the representation of the impact of light intensity (attenuated by culture media depth) and temperature on the photoautotrophic growth of *Cyanothece* [9] (and subsequent scale-up to pilot scale for biomass production [10], [11]);
- the optimisation of nutrient removal by microalgae, calibrated and validated with both respirometric and titrimetric data [10];
- the construction and experimental validation of a model of microalgae biomass and lipids accumulation in a PBR using initial TN, light intensity and temperature as the primary inputs [11];
- a simple model to predict biomass values, using kinetic growth parameters, as a function of HRT to maximise biomass productivity in a continuous PBR [12].

Despite the large number of mathematical models presented in the literature to simulate PBRs, no models have been proposed which comprehensively incorporate all phenomena relevant to combined biomass growth and nutrient removal from wastewater. Most of the published models, including the most recent [10], do not address the effects of the initial nutrient concentration on both the treated wastewater quality and algal growth. Whilst the classic mathematical dynamic model of microalgae growth proposed by Droop [13, 14] accounts for the dilution rate and effects of inorganic nitrogen concentration, the impact of all other parameters (including CO₂ gas concentration $C_{c,g}$, light irradiation intensity I , and the nutrient uptake rate) is neglected. Models have otherwise been developed for lipid synthesis as it relates to algal growth rate and N uptake [15], and luxury uptake of phosphorus as polyphosphate as a function of the available P, light intensity and temperature so as to provide P removal efficiency [16]. Thus, whilst a wide range of base parameter values that have been employed in these models (Table 1), these have not led to the same sort of treated effluent nutrient concentration profile outputs as provided by classical biological nutrient removal (BNR) models [17].

Table 1: Parameter values assumed in recently-published PBR mathematical models, batch processes

<i>PBR configuration</i>	μ_{max} d^{-1}	k_d d^{-1}	$C_{c,g}$ %v/v	Q_g vvm	HRT d^{-1}	$K_{L,C}$ d^{-1}	TP, mg L^{-1}	TN, mg L^{-1}	TC, mg L^{-1}	T , C°	I , $\mu E m^{-2}$ s^{-1}	<i>Ref.</i>
Classical PBR	0.1- 0.52	--	0.03 ^a	0.5	3.3 ^s	--	12.7	54.5 8	384 ^c	20±3	90	[8]
Tubular PBR	1.75	--	10	0.028	1.36 ^b	--	--	--	--	25-37	275-23	[9]
Flat plate PBR	0.156	--	--	--	1- 10 ^b	--	--	--	--	23	150	[18]
Bubbled PBR	--	--	0.03 ^a	--	280	--	1.12	12	--	25	250-30	[19]
Breeding reactor	0.1-2	--	0.03 ^a	--	--	5-10	4	6	--	26	130-90	[10]
Combine algal unit	0.1-11	--	2	2.5	--	0.3-19	--	--	--	20	90	[20]
Solix PBR	0.6	0.01	2.5	--	65	--	1.12	12	--	21-24	200	[11]
Helical PBR	1.77	--	4	0.047	15- 2.5 ^b	--	5	260	5	24-33	156	[21]
BIOSTAT PBR	0.52	--	10	16	2.5	--	--	--	--	25	--	[22]
Flat panel PBR	0.94	--	5	0.63	3.4- 1.1 ^b	1.63	31	--	--	20	250	[12]

μ_{max} , maximum specific growth rate; k_d , biomass loss rates; Q_g , gas flow rate; vvm = volume gas per volume liquid per minute; HRT hydraulic residence time; $K_{L,C}$ CO₂ mass transfer rate; TP total phosphorus; TN total nitrogen; TC total carbon, I irradiation intensity
^aAtmospheric level; ^bContinuous or ^ssemi continuous system; ^cCOD.

There is an evident need for a numerical mathematical model capable of simulating algal growth rates as a function of initial TP, TN including the influence of $C_{c,g}$ and I . Such a model could then be employed to estimate nutrient removal and predicated dynamic behavior of a batch system.

2. Materials and methods

2.1. Practical experiments

The work was based on *Chlorella vulgaris* Cv (strain: CCAP 211/11B, CS-42), supplied by Australian National Algae Culture Collection/CSIRO Microalgae, and inoculated with a 10% suspension of microbiology-derived medium (MLA) [23]. The cells were cultivated in a shaker incubator at 200 rpm and 20°C, with continuous illumination of white fluorescent light at 50 μE for three weeks inside an incubator refrigerator.

Experiments were conducted in 350 mL cylindrical glass columns (ID = 4 cm) with a 250 mL working volume. Sterilized 250 mL medium at different concentrations of TN (0 - 207 mg L⁻¹) and TP (0-8 mg/L), was inoculated with 1 vol% pre-cultured medium. The culture was fed continuously with filtered air, enriched with CO₂ (0.04-5%), at a flow rate of 50 mL min⁻¹ as monitored by digital mass flow controllers (MC-100SCM, Cole-Parmer, USA); the inlet and outlet CO₂ concentration was measured using a CO₂ probable meter (G110, Geotech, UK). Continuous illumination between 100 and 250 μE m⁻² s⁻¹, provided by adjusting the number of 8W LED lights between 4 and 8, was measured by a light meter (LI-250A, LI-COR, US & Canada). A 5 mL sample was extracted daily for analysis, equating to a hydraulic and solids residence time of 50 days, and all runs lasted for 10-13 days.

The total nitrogen (TN) and total phosphorus (TP) concentrations of the filtered liquid sample (with membrane filter, 0.45 μm) were analysed using a HACH kit (DR/890 Colorimeter, HACH, USA) and the total organic carbon (TOC) concentration using a Shimadzu TOC analyser. The optical density (OD) was determined by UV-Vis spectrophotometer (Jasco V - 670) at 680 nm, and the reading converted to dry cell weight ($X \text{ g L}^{-1}$) by calibration. The specific growth rate μ was then calculated from the initial and final biomass concentrations and the corresponding cultivation time. For all nutrient tests the control sample contained 6 mg L^{-1} TP and 28 mg L^{-1} TN, based on the typical medium MLA composition stipulated by the supplier.

2.2. Modelling equations

The model development proceeded through the steps indicated in Figure 1. The mathematical model used to simulate the experimental data in this study was based on the classical homogenous model for a bubble column PBR operated in batch mode [21]. The model was developed as a set of parameterized nonlinear first order differential equations defined by fundamental physical and/or chemical mechanisms and base experimental data. Biological, gas and liquid phases considered in the PBR and mass balance equations were derived according to the following assumptions:

- All cultivation conditions factors affecting C_V growth were encompassed, including gas flow rate, light intensity, temperature, feed water quality.
- The Henry constant (H) was considered to be unaffected by the gas pressure: values were taken at 25°C and atmospheric pressure for both CO_2 and O_2 .
- Microalgal cells were assumed to be able to fix dissolved inorganic carbon regardless of its form (i.e. CO_2 , HCO_3^- and CO_3^{2-}).
- The light intensity was sufficient to ensure growth without being impaired by dissipation by the biomass concentration but not so excessive so as to cause light limitation, with the threshold biomass concentration assumed to be 1 g L^{-1} [2].
- Light inhibition (due to excessive irradiation levels of 500-2500 $\mu\text{E m}^{-2} \text{ s}^{-1}$ [24, 25]) was ignored; a range of 100-250 $\mu\text{E m}^{-2} \text{ s}^{-1}$ light irradiation range was employed.
- Operation was assumed not to be limited by the nutrient or carbon concentration.

Some simplifying assumptions were made in developing the model to reduce the complexity of the model parameters:

- The gas flow rate was considered to remain constant with time and culture depth: the gas was assumed to be uniformly distributed in the cultivation medium.
- Oxygen inhibition was ignored, since it was removed to low levels in the experimental tests.
- A batch growth culture was considered, although the model could be readily adapted to a semi-batch or fed-batch reactor.
- Only autotrophic growth from light irradiation was considered: other possible growth mechanisms were ignored.

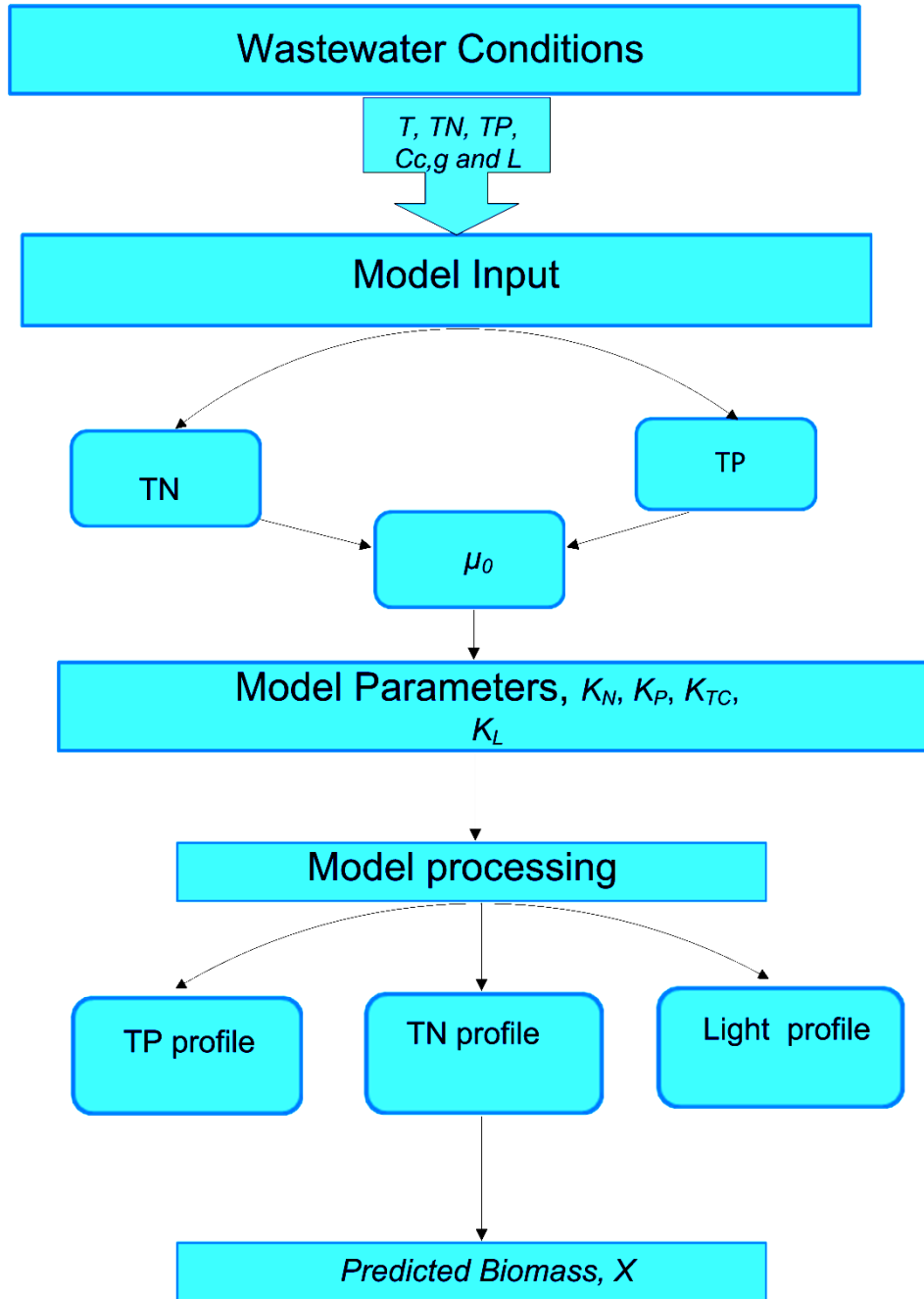


Figure 1. Modelling strategy

2.3. Gases dynamics in liquid phase of PBRs

The rate of CO₂ transferred from the gas to liquid phase is represented by dual-film theory:

$$R_C = k_c(C_s - C)V(1 - \varepsilon) \quad (5)$$

Where k_C is the mass transfer coefficient for the transfer of CO_2 from the gas phase to bulk culture phase, C_s is the saturated concentration of CO_2 , C is the concentration of inorganic carbon, V the reactor volume and ε the gas hold up volume. According to Henry's law:

$$C_s = \frac{Py}{RTH} \text{MW}_{\text{HCO}_3} 1000 \quad (6)$$

where P is the pressure, y the gas phase CO_2 fraction, R the gas constant, T the temperature, H the Henry's Law constant and MW_{HCO_3} the molar mass of bicarbonate. k_C can be estimated from correlations available [26] for the transfer coefficient for oxygen k_O using the aqueous phase diffusivities of CO_2 and O_2 (D_C and D_O respectively):

$$k_C = k_O \sqrt{\frac{D_C}{D_O}} \quad (7)$$

ε was estimated by volumetric expansion as proposed by Chisti [27] based on the gassed and un-gassed height of fluid (h_G and h_L respectively) in each part of the PBR:

$$\varepsilon = \frac{h_G - h_L}{h_G} \quad (8)$$

The uptake rate of carbon by the algal culture can be expressed as a function of biomass concentration, the yield coefficient (total carbon consumed per algal biomass produced $Y_{C,tot}$) and μ_X , the algal specific growth rate:

$$\frac{dc}{dt} = -1000 Y_{C,tot} \mu_X X \quad (9)$$

The total carbon dissolved in the algal culture can be obtained from substrate equations (5) and (9).

$$\frac{dC_{total}}{dt} = k_C (C_s - C) V (1 - \varepsilon) - 1000 Y_{C,tot} \mu_X X \quad (10)$$

with appropriate initial conditions of:

$$[dC_{total}] = [C_{init,total}] @ t=0 \quad (11)$$

The first term on the right hand side of Equation 10 takes into account mass transfer phenomena from the gas to liquid phase and the second term the CO_2 consumption and concomitant liquid phase microalgae production process.

The mass balance for total dissolved nutrients (N and P) not involved in the gas liquid mass transfer phenomena can be expressed as follows:

$$\frac{dC_{N,P}}{dt} = -1000Y_{N,P} \mu_X X \quad (12)$$

with appropriate initial conditions:

$$[N, P] = [N_{init}, P_{init}] \quad @ \quad t=0 \quad (13)$$

2.4. Biomass growth rate

The final growth rate of algal biomass, dX/dt , can be written as :

$$\frac{dX}{dt} = \mu_X X - K_d X \quad (14)$$

where K_d is the biomass loss rate. According to Monod model the nutrient-limited algal growth rate μ_0 can be expressed as

$$\mu_0 = \mu_X \left[\frac{S_i}{K_i + S_i} \right] \quad (15)$$

where S_i is the nutrient (N or P) concentration, K_i the nutrient half saturation constant and μ_X , the non-nutrient limited specific growth rate, is determined experimentally.

Specific experiments were carried out by cultivating *Cv* in batch PBRs at various initial dissolved nutrient concentrations to establish the impact on growth. Outputs were used in conjunction with Equations (16-17) to evaluate the dependency of μ_0 on initial dissolved nutrient concentration:

$$\mu_P = 0.989 \left[\frac{S_P}{1.65 + S_P} \right] \quad (16)$$

$$\mu_N = 1.039 \left[\frac{S_N}{8.7 + S_N} \right] \quad (17)$$

where the values of 0.989, 1.039, 1.65 and 8.7 are the Monod parameters calculated from the baseline experimental data (Section 3.2). Overall, for the combined effect of N and P nutrients on the algal growth, the Double Monod model [28]:

$$\mu_{N,P} = \mu_{max} \left[\frac{S_N}{K_N + S_N} \right] \left[\frac{S_P}{K_P + S_P} \right] \quad (18)$$

where S_N , S_P are the respective N and P concentrations in the culture and K_N , K_P the corresponding half saturation constants for nitrogen and phosphorus respectively, based on the experimental data. The specific growth rate is then given by:

$$\mu_0 = 1.4 \left[\frac{S_N}{6.5 + S_N} \right] \left[\frac{S_P}{0.75 + S_P} \right] \quad (19)$$

where the values of 1.4, 6.5 and 0.75 are the Double Monod model parameters again calculated from the baseline experimental data (Section 3.2). Extending this equation to account for growth control substrate for both N and P:

$$\mu_x = \mu_0 \left[\frac{S_N}{K_N + S_N} \right] \left[\frac{S_P}{K_P + S_P} \right] \left[\frac{TC_c}{K_{TC} + C_c} \right] \left[\frac{L_{av}^n}{K_L^n + L_{av}^n} \right] \quad (20)$$

where C_c is the total carbon dissolved in the culture media, K_{TC} total carbon half saturation constant and K_L is half saturation constant of light. The average light intensity L_{av} within the culture in cylindrical bubble PBR illuminated by a unidirectional parallel flux can be expressed by:

$$L_{av} = \frac{2I}{r K X \pi} \left[1 - \int_0^{\pi/2} \cos(\theta) \exp(-2rKX \cos(\theta)) d\theta \right] \quad (21)$$

where r represents the PBR radius. K is the overall light extinction coefficient, given by:

$$K = K_c C_{chl} + K_w \quad (22)$$

where K_c is the chlorophyll-based light extinction coefficient of algae, C_{chl} is the chlorophyll concentration, which is a function of biomass concentration and is determined experimentally, and K_w is the light extinction coefficient of pure water.

The combined set of ordinary differential equations (*ODEs*) were coded in MATLAB to correlate the time-dependent algal biomass concentration X (g L^{-1}) and algal nutrient uptake functions with light intensity, and the simulations validated with the experimentally-determined C_v growth data. A sensitivity analysis of a parameter P_j was conducted with respect to X to assess the response of biomass concentration to changes in each model parameter to assess the validity of the simplifying assumptions (Section 2.2):

$$\sigma_x = \frac{\Delta X}{X} \frac{P_{jnom}}{\Delta P_j X_{nom}} \quad (23)$$

where P_{jnom} is the parameter nominal value and X_{nom} the model response when nominal parameters are used. A $\pm 20\%$ variation in ΔP_j was applied to obtain the test values to

determine ΔX . Four biomass profiles were used in calculating the mean profile with the standard deviation estimated from the four runs. The sensitivity coefficient for each parameter was calculated from the average spread according to the method of Bernard et al. (2001) [29].

An F-test was performed to determine the variance between the predicted and measured values using *Jmp* statistical discovery software (*SAS version 11.2.1*).

3. Results and discussion

3.1. Mass transfer coefficient:

k_C was estimated from k_O data using Equation 7. A previous reported correlation for bubble columns [30, 31] was used to calculate k_C in terms of ε measured at the corresponding gas flowrate. This procedure produced a k_C of $5.8 \times 10^{-6} \text{ s}^{-1}$ at a gas flowrate 50 mL min^{-1} , towards the low end of the range of values reported in the literature of $3.8 \times 10^{-6} \text{ s}^{-1}$ [20] for different PBR configurations.

3.2. Model calibration

The influence of the parameters TN_{init} , TP_{init} , I and $C_{c,g}$ on the kinetics of C_V was determined through mathematical simulation, and the reliability of the model data assessed using validatory experimental test data. Further experiments were carried out to evaluate the influence of the initial concentrations of dissolved inorganic carbon, N and P on algal growth by varying the concentrations of sodium bicarbonate, sodium nitrate and potassium bisphosphate in the MLA cultivation medium. The feed gas $\underline{C}_{c,g}$ was also varied at a constant Q_g of 50 mL min^{-1} . A pH 6-8 was maintained throughout.

Calibration was through determination of μ_{max} from Equations 20-22 tuned through a nonlinear fitting procedure. The computed biomass concentration values were compared by the least squares method with microalgal growth experimental data. The relative error obtained by the fitting procedure was around 1%. Other parameters were estimated, such as the half saturation constants for light (K_L), carbon (K_{TC}), and nutrients (K_N and K_P), the algae loss rate (K_d), the chlorophyll-based light extinction coefficient (K_C), the light extinction coefficient of pure water (K_w) and the shape parameter (n) were then adjusted by trial and error to obtain the best fit for the biomass concentration profile. The best fit parameters established by calibration were used in the validation step based on the growth profile data from the remaining experimental tests runs. Biomass yield values Y_i ($\text{g}_w \text{ g}_{biomass}^{-1}$) were estimated according to published methods [32] from the C_V elemental composition, assumed to have the formula $C_{1000}H_{1650}O_{242}N_{130}P_{10}S_{4.5}K_{2.7}Na_{2.5}Mg_2Ca_1Cl_{0.2}$ [33].

Values for the shape factor parameter n , accounting for light limitation caused by the biomass density, were adjusted according to the assumptions presented in Section 6.1. The best-fit value obtained for n was 1.4, this value being slightly outside the range of 1.49-2.2 reported in previous studies [21, 28, 34]. Best-fit values generated by simulation during the calibration process for other parameters are shown in Table 2.

Table 2: Summary of base parameters values

<i>Symbol</i>	<i>Unit</i>	<i>Parameter value</i>	<i>Notes</i>
μ_{max}	s^{-1}	1.62×10^{-5}	Current study, estimated
K_d	s^{-1}	2.3×10^{-6}	Current study, estimated
K_L	$\mu E m^{-2} s^{-1}$	16	Current study, estimated
K	$m^2 g^{-1}$	2.5×10^{-1}	Current study, estimated
K_p	$g L^{-1}$	4×10^{-2}	Current study, estimated
K_N	$g L^{-1}$	3.7×10^{-1}	Current study, estimated
K_{TC}	$g L^{-1}$	5×10^{-1}	Current study, estimated
N	-	1.42	Current study, estimated
$H_{e,C}$	-	8.32×10^{-1}	[35]
$H_{e,O}$	-	3.2×10^{-2}	[35]
Y_{ctot}	$g_c g_{biomass}^{-1}$	5×10^{-1}	Current Study, estimated
Y_p	$g_P g_{biomass}^{-1}$	2.4×10^{-2}	Current Study, estimated
Y_N	$g_N g_{biomass}^{-1}$	2×10^{-1}	Current Study, estimated
Y_O	$g_{O_2} g_{biomass}^{-1}$	0.534	(--), [36, 37]
d_B	m	1×10^{-6}	Current Study, estimated
D_C	$m^2 s^{-1}$	14.7×10^{-9}	[30]
D_O	$m^2 s^{-1}$	8.0×10^{-9}	[30]
$K_{La,C}$	s^{-1}	5.78×10^{-6}	Current Study, estimated
K_w	cm^{-1}	0.0018	[38]

3.3. Model validation

The model was validated by comparing the experimentally-determined biomass concentration profiles for the remaining five sets of experimental conditions with model predications using the parameter values determined by the calibration process. According to growth profile data for a range of different initial nutrient concentrations ($TN_{init} = 28-207 \text{ mg L}^{-1}$; $TP_{init} = 6-8 \text{ mg L}^{-1}$) and gaseous carbon concentrations ($C_{c,g} = 0.03-5 \%$), there is a reasonable fit between the experimental and model data (Fig. 2). The base parameters identified (Table 2) were thus used for subsequent sensitivity analysis (Section 3.4) and nutrient removal profile modelling (Section 3.5).

3.4. Sensitivity and regression analysis

The sensitivity of the growth profile to the parameters L_k , K_N , TC_K , K_p , K_d and μ_{max} was examined based on the experimental conditions of $C_{c,g} = 2.5\%$, $TN_{init} = 28 \text{ mg L}^{-1}$, $TP_{init} = 6 \text{ mg L}^{-1}$ and $I = 250 \mu E m^{-2} s^{-1}$ used for the calibration step. The pre-defined model constants (Table 1) were used as base values and individually varied by $\pm 20\%$ and the impact on X . The mean predicted profile from four runs for each parameter is shown in Fig. 3 with the corresponding sensitivity coefficient estimated from Equation 23, with sensitivity increasing with increasing σ_x . According to Figure 3, μ_{max} ($\sigma_x = 0.52$) and the biomass loss rate (K_d , $\sigma_x = 0.11$) were the most sensitive parameters followed by considerably reduced sensitivity for the half saturation constant for carbon and light at ($\sigma_x = 0.008$ and 0.004 for K_{TC} and K_L respectively). Sensitivity to μ_{max} has been previously reported [39] and reflects the importance of this parameter on the accuracy of model prediction. The influence of remaining four parameters on biomass profile was negligible.

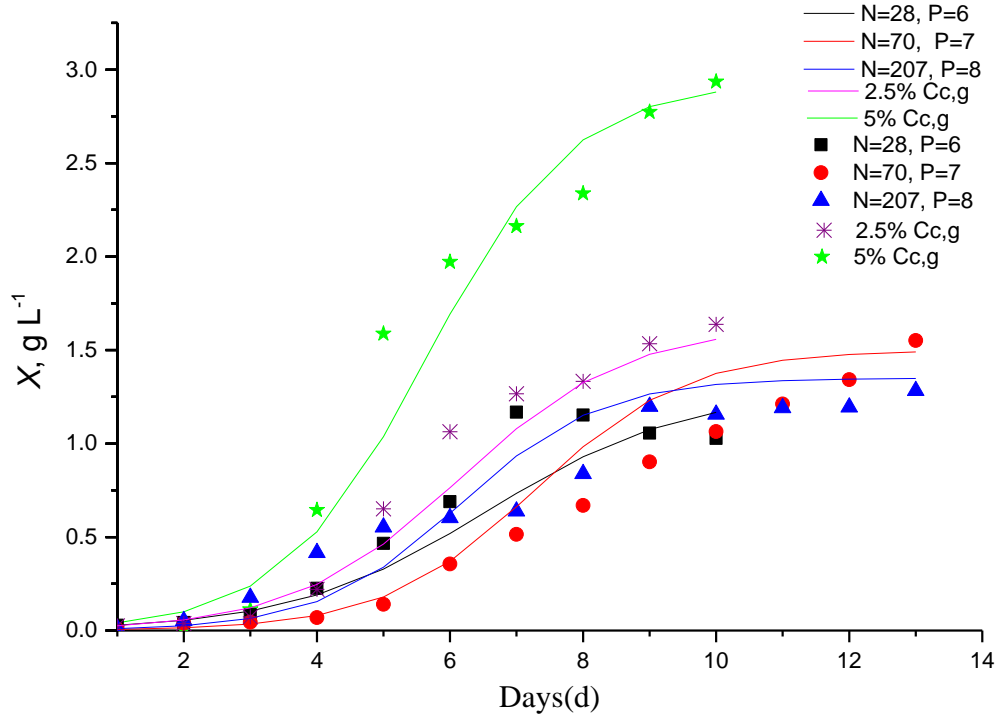


Figure 2. Model verification applying different cultivation conditions (TN_b , TP_b , $C_{c,g}$ and I for experimental (data points) and model (continuous trend) data

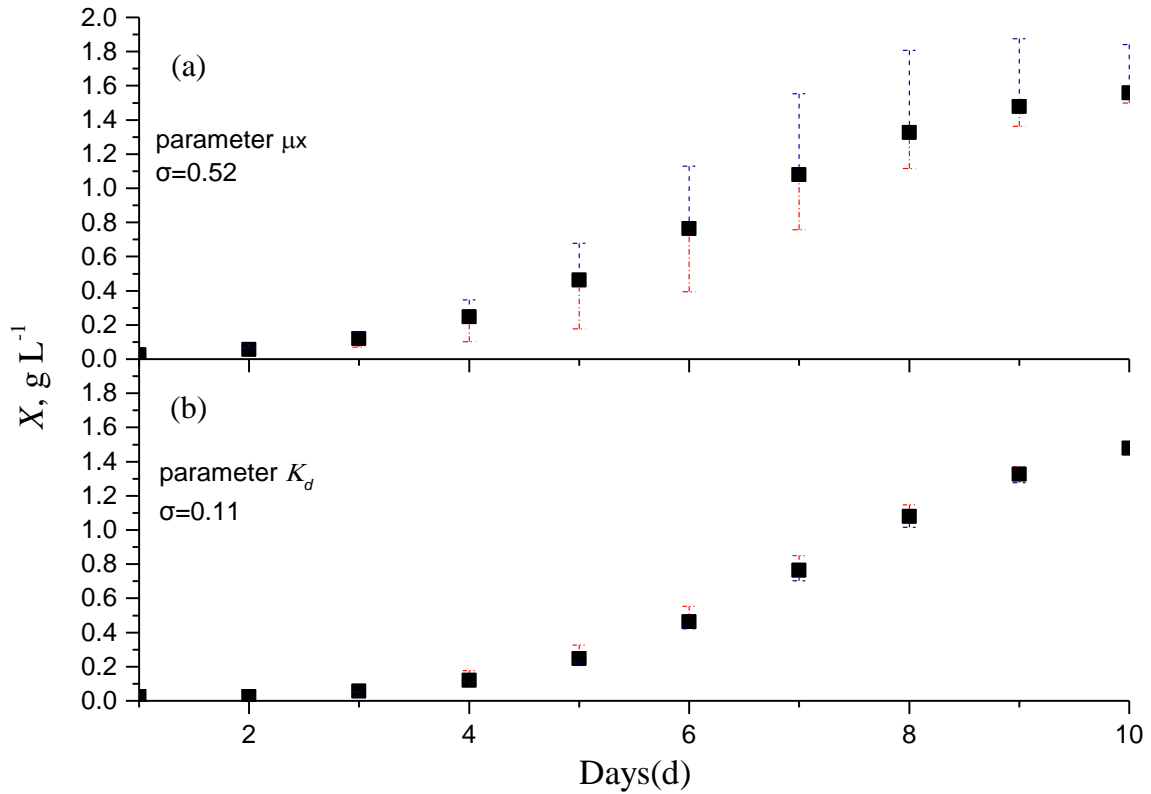


Figure 3. Predicted mean algal biomass concentration profile and its variation based on $\pm 20\%$ variation in (a) μ_x , and (b) K_d , based on from four runs.

The model was further validated by applying the regression analysis for the experimental and predicted growth profiles using SAS software. Significance of the fits achieved between the measured and predicted data (Fig. 4) is illustrated by R^2 values of 0.94 and p values below 0.0001 for the all validated points. The model appears to reflect the growth dynamics at various values of TN_{init} , TP_{init} , I and $C_{c,g}$ with reasonable accuracy, comparable to that reported by Wang et al [40] who investigated the kinetics of nutrient removal and characterised the extracellular polymeric substances (EPS) generated. The specific growth rate μ_{max} calculated from Eq. 18 was found to be the most sensitive parameter .

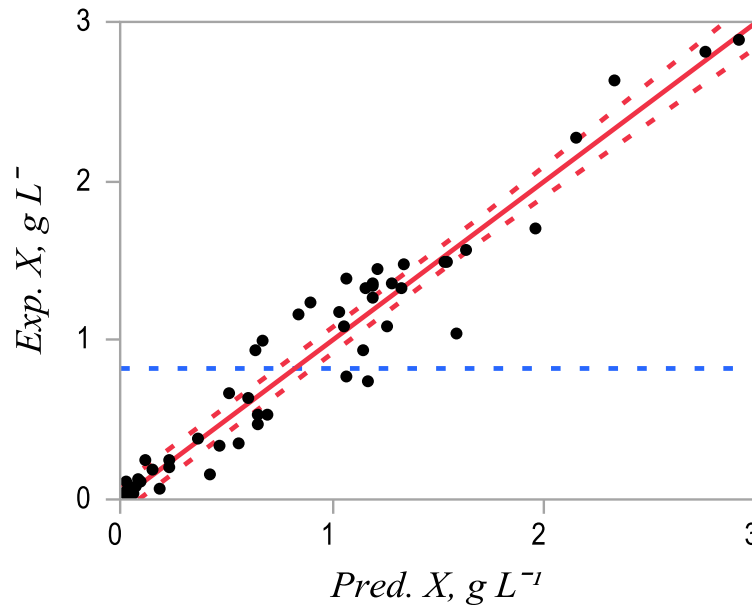


Figure 4. Regression analysis of fit between predicted and experimental biomass concentration $R^2 = 0.94$, $P = 0.0001$ and $p > F = < 0.0001$ (i.e. significant). Dotted curved lines indicate $>95\%$ confidence bands; horizontal dotted lines represent mean of the Y leverage residuals (i.e. the measure of agreement with the model).

3.5. Nutrient removal

The evolution of nutrient uptake and algal growth kinetics as a function of TN_{init} and TP_{init} was modelled and the outputs compared with experimental data. Modelled outputs were based on Equation 12 and encompassed the yield coefficient for total N or P ($Y_{N,P}$), μ_x and biomass concentration X , each of these parameters directly affecting the nutrient uptake rate.

Nutrient consumption transients (Fig. 5a-d) indicated rapid removal of P to 100% removal, such that it becomes limited by around 6 days at the lowest TP_{init} of 6 mg L^{-1} (Fig. 5a). Whilst the extracellular phosphorous is depleted rapidly in the cultivation medium, the cells continued to grow over the period of the experiment (Fig. 5a). The luxury uptake of nutrients and storage for later growth is a well-established phenomenon in phytoplankton [14], although this does not influence the P uptake rate. TN removal efficiencies of 80-99% were recorded after 10-13 days for TN_{init} concentrations up to 70

mg L⁻¹, whilst only 60% of the TN was removed after 13 days at the highest TN_{init} employed of 207 mg L⁻¹ (Fig. 5c). The model appears to adequately predict the dynamic depletion of TN and TP in the cultivation medium, along with algal biomass production (Fig. 4), over the ranges of initial nutrient concentration (Figs. 5a,c) and feed gas CO₂ concentration (Figs. 5b,d) studied.

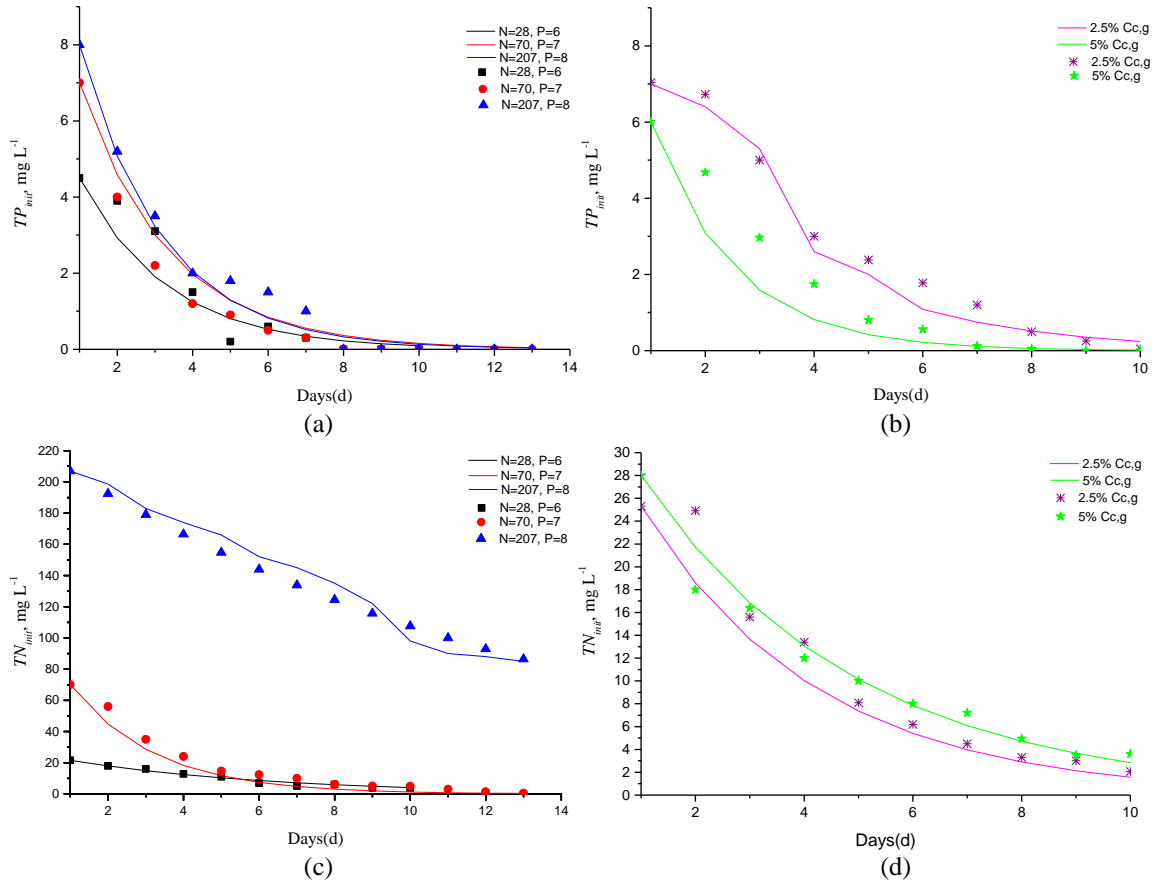


Figure 5. Evolution of total nutrient concentration under different conditions of initial aqueous nutrient concentration (a, c: $C_{c,g} = 0.03\%$) and feed gas CO₂ concentration (b, d: TP_{init} and $TN_{init} = 6$ and 28 mg L⁻¹ respectively)

3.6. Light attenuation profile

The predicted average irradiance I_{av} profiles over the course of the test period are shown in Figure 6. In all cases examined it was confirmed that the limiting step for autotrophic PBR operation was the availability of light energy: the maximum productivity is dictated by rate at which the light can be absorbed and transformed into biomass.

There was a reduction in biomass concentration (of 1.6 and 2.9 g L⁻¹) when the microalgae cells were illuminated at 200 and 250 $\mu E m^{-2} s^{-1}$ and fed with 5 and 2.5% CO₂ gas respectively. Evidence therefore suggests that for irradiance values above the light saturation point photo-oxidation takes place, damaging the photosystem and inhibiting photosynthesis and microalgae growth [41]. Under such circumstances, the photo-inhibition rate (which takes place at all irradiance values) exceeds the rate of repair of the

algal cells [42]. However, below the light saturation point the expected proportional increase in biomass productivity and in CO₂ uptake, from photosynthesis, with increasing light irradiance was observed. The photosynthetic rate is greater at lower biomass concentrations due to the increased light availability associated with the higher light transmission through the suspension.

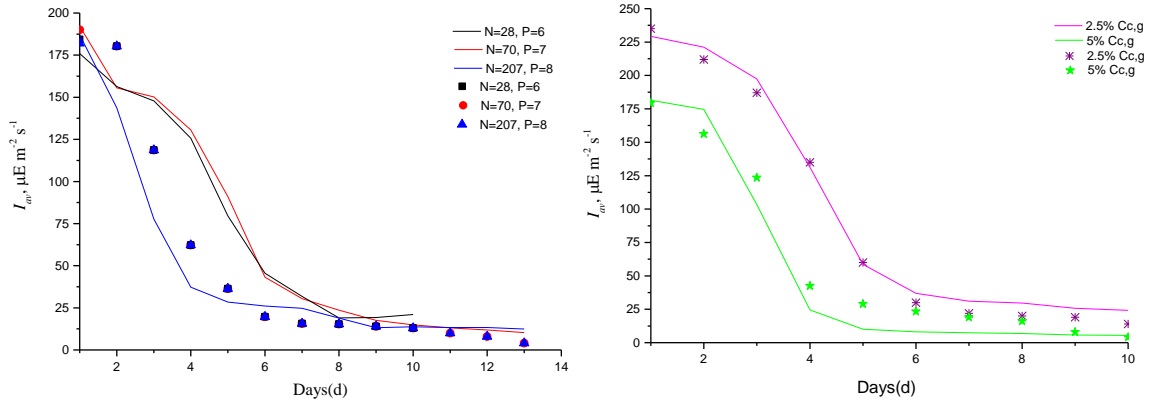


Figure 6. Average irradiance I_{av} profile during experiment period of 10 days in dashed line and the predicated response in solid line.

The proposed model (combining growth and light profile) was fitted to experimental data conducted under different light intensities utilizing only a single set of parameters. The light model parameters, the half saturation constant and overall extinction coefficient were estimated experimentally as before (Table 1), with the parameter estimation based on non-linear least square fitting of the light model-predicted data to that derived from experiments conducted at different light intensities. These parameters were then numerically optimized to obtain a best fit of the model to experiment data using the simplex search algorithm of MATLAB [43]. The term n (Eq.20) takes account the abruptness of the curve in the transition from low to high irradiance. The half saturation constant for photosynthesis, L_k , was determined experimentally by measuring the kinetic growth constant under different light intensities. Since the average irradiance at the end of the cultivation period was greater than the half saturation constant for light I_k of $16 \mu E m^{-2} s^{-1}$, the cultures were not predominantly photo-limited and the non-linear behavior response to the light attenuation caused by absorption and scattering associated with the biomass increase during the growth phase.

4. Conclusions

The dynamic behavior of an algal photobioreactor (PBR) has been modelled with reference to both biomass growth and nutrient uptake and associated removal from a water stream. The model was calibrated through using a base set of experimental conditions of feed gas CO₂ concentration ($C_{c,g} = 2.5\%$), initial nutrient concentration (TN_{init} and $TP_{init} = 28$ and 6 mg L^{-1} respectively), and an irradiation intensity (I) of $250 \mu E m^{-2} s^{-1}$.

The model outputs were found to be sensitive only to the maximum specific growth rate (μ_{max}) and the biomass loss rate K_d , with a possible maximum $\pm 46\%$ deviation in the computed algal biomass concentration (X) associated with a $\pm 20\%$ variation in μ_{max} . The calibrated model was otherwise able to predict the biomass concentration X with reasonable accuracy: regression analysis revealed a correlation coefficient (R^2) of 0.94 for mathematically-predicted vs. experimentally-determined values of X over a wide range of input parameter values of feed gas and initial nutrient concentrations. Dynamic nutrient removal was similarly accurately predicted as reflected in the corresponding R^2 values of 1.00 and 0.95 for the N and P profiles respectively. Predicted values based on the determined average light profile were slightly less accurate, with R^2 values of 0.91.

The calibrated model has been successfully demonstrated for the prediction of algal growth and nutrient removal, encompassing the prediction of the light profile within the biomass bulk as a function of biomass concentration. The latter would be expected to inform the appropriate dilution ratio in a continuous process. However, further work is required to determine how far the modelling approach can be extended to other operating conditions, reactor configurations and algal species.

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