Continuous Algae-based Carbon Capture and Utilization (CACCU) to Transform Economics and Environmental Impacts: DE FE 0032108

Texas A&M University Washington University in St Louis NCCC at Southern Company



U.S. Department of Energy National Energy Technology Laboratory Carbon Management Project Review Meeting

Project managed by Dr. Lei Hong

Susie Dai Texas A&M University August 08, 2024



Project Overview

- Funding
 - DOE \$2,000,000; Cost Share \$510,583
- Overall Project Performance Dates
 - Original 10/01/2021-09/30/2024, three BPs
 - Currently: with 12-month extension to Sep 2025, in BP3
- Project Participants
 - TAMU: Drs. Susie Dai, Joe Zhou, Bruce McCarl, Stratos Pistikopoulos, Chengcheng Fei
 - WUSTL: Drs. Young-shin Jun, Yinjie Tang, Joshua Yuan, Benjamen Kumfer
 - NCCC at Southern Company: Frank Morten, Tony Wu

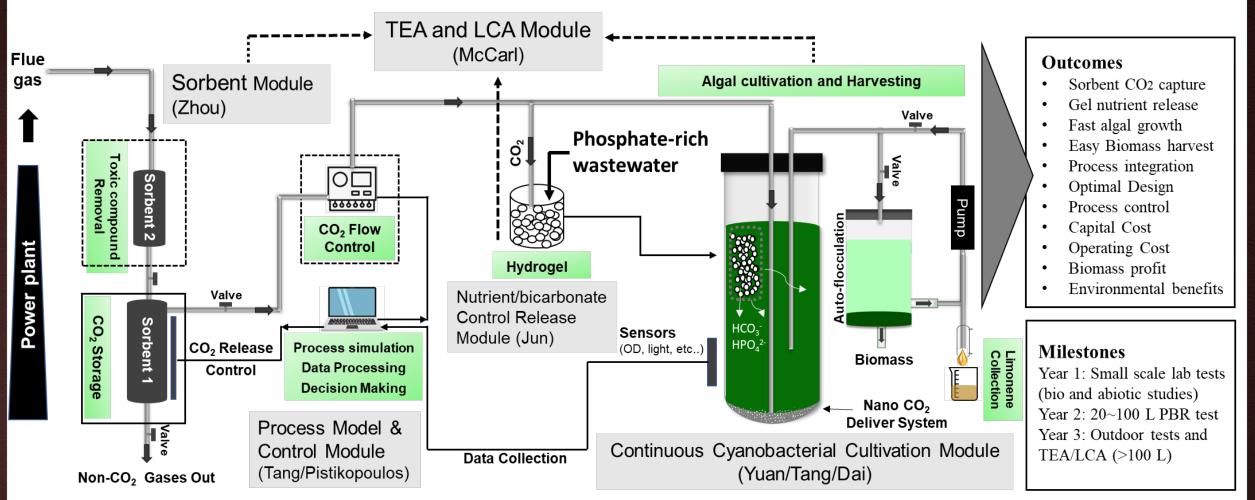


Overall Project Objectives

- The project integrates novel CO₂ capture/controlled release sorbent with a breakthrough continuous algal cultivation system, assisted by hydrogel technology to reduce media cost, fertilize the algae with controlled nutrient delivery.
- Objective 1: Project management.
- Objective 2: Integrates CO₂, bicarbonate, and nutrient capture and delivery to the low-cost harvest-empowered continuous algal cultivation system with ultra-high productivity and CO₂ uptake plus valuable chemical bioproduct production. We also advance algal strain, sorbent, and hydrogel technologies to enhance carbon capture and yields of limonene, biomass, and glycogen.
- Objective 3: Scale up the sorbent technology and integrate it with algal cultivation.
- Objective 4: Test the prototype CACCU system with flue-gas coupled 100 L photobioreactor (PBRs).



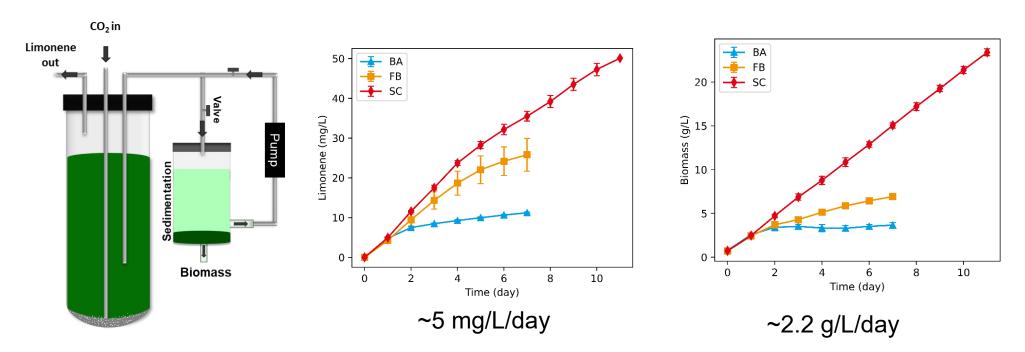
Technology Background



The integrated CACCU system



Sustainable co-production of limonene and biomass by semicontinuous cultivation



Dai and Yuan's groups@TAMU

Record productivities and yields in limonene productivity

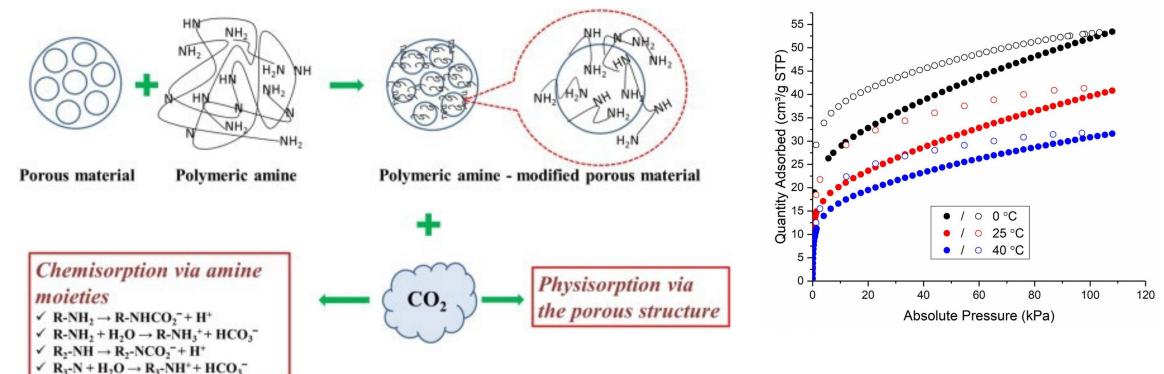
Sustainable biomass accumulation at about 1-2g/L/Day for a long period of time.

Machine learning informed semi- continuous cultivation.



Long et al., Nature Communications, 2022, 13:541

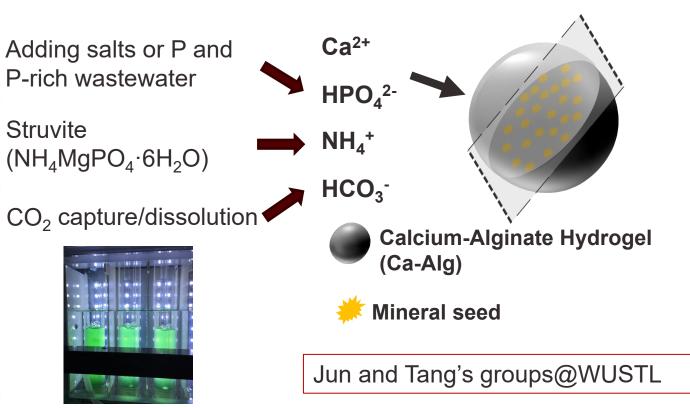
Amine Grafted Porous Polymer Network

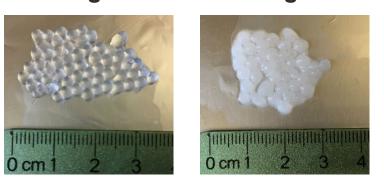


CO₂ adsorption of PPN-151-DETA



Mineral-seeded mineral hydrogel composites for nutrient delivery and pH control Ca-Alg Ca-Alg/CaP





Ca-Alg/CaCO₃ Ca-Alg/CaP+CaCO₃ Ca-Alg/CaP+CaCO₃

Kim, D and Jun, Y.-S., Green Chemistry 2018, 20 (2), 534-543.

- Calcium phosphate, calcium carbonate, or ammonia-containing mineral seeds formed during alginate crosslinking.
- When placed into calcium phosphate/carbonate supersaturated solution, mineral seeds grow, collecting and incorporating phosphate, bicarbonate, and ammonia-containing minerals.



Translating process models into a process systems engineering framework at scale involves some critical steps

- 1) Accurate modeling of process dynamics
- 2) Reduced order approximation of nonlinear dynamics
- Surrogate linear models can tame computational complexity
- Linear programs can provide certificates of optimality
- 3) Design of control scheme

4) Formulation of a network design as a mixed integer program (MIP)

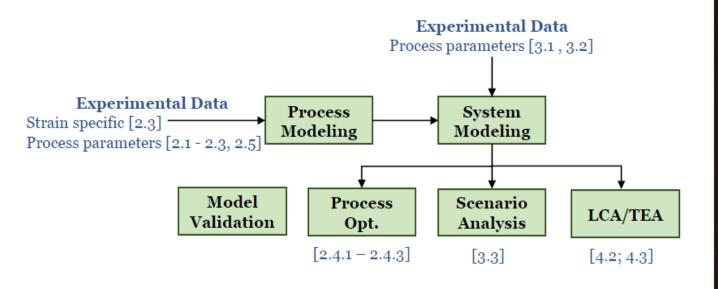
MIPs can be optimized to multiple objectives

Network decisions can be modeled as binary variables

Scheduling can be integrated (multiscale approach)

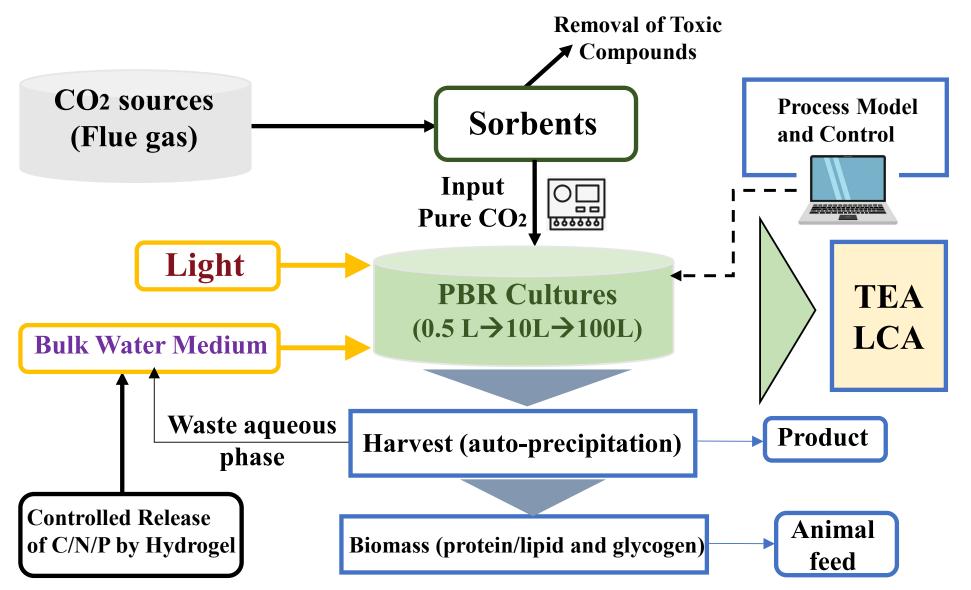
5) Integration of lifecycle tools

OpenLCA data integration with MIP framework (MIP)



Tang, Pistikopoulos and McCarl's groups@TAMU&WUSTL

Technical Approach/Project Scope

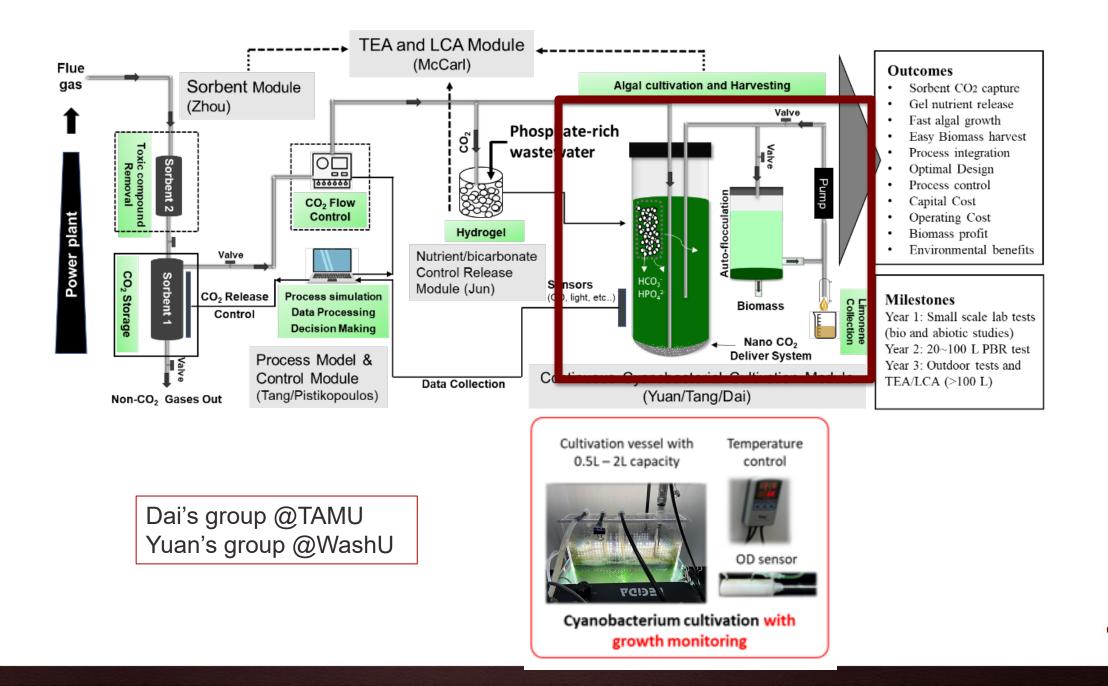




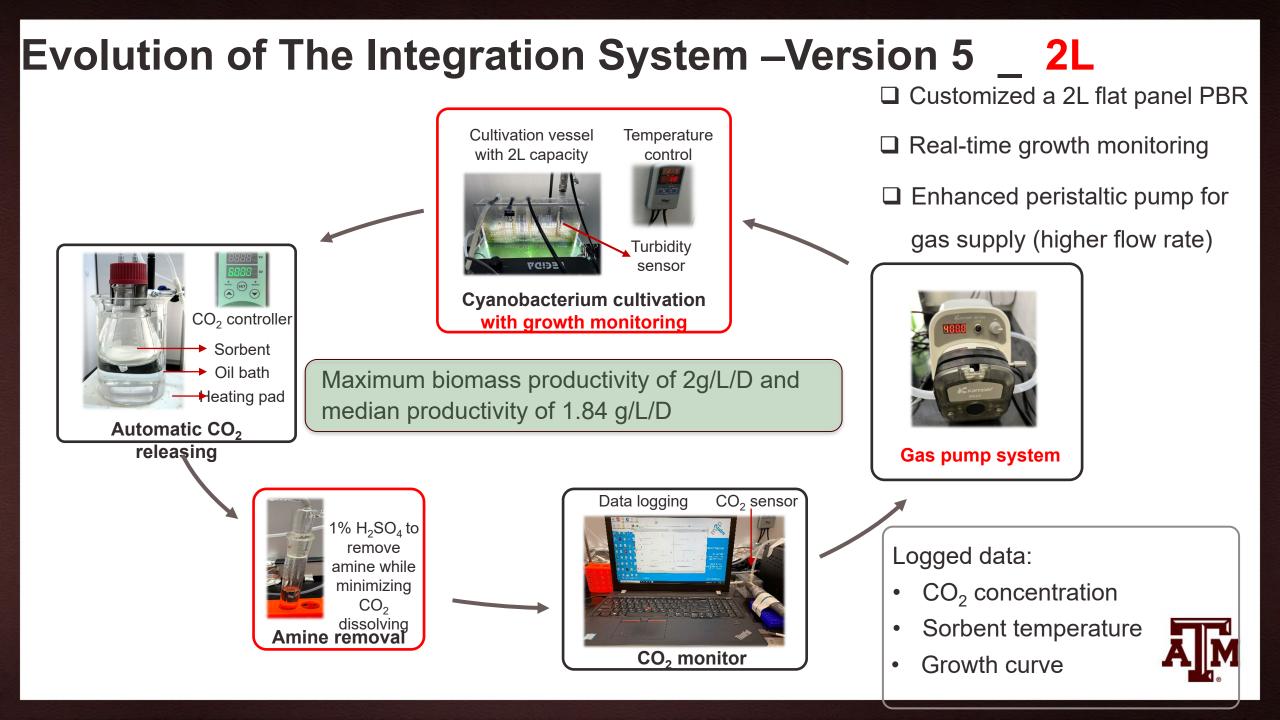
Progress Tasks to be finished in BP2

Task	Milestone Title	Complete Date	Status
2.4.2	Compare the process control model and machine learning model and decide the strength and applicability of each model.	Q10(03/31/2024)	Completed Build the MAGMA model and plan to submit the manuscript in 2024.
2.5	Scale up to bench-scale at 2 Liter with algae biomass yield >1.8 g/L/D using engineered strains and sorbent-released CO_2	Q10(03/31/2024)	Completed Maximum biomass productivity of 2g/L/D and median productivity of 1.84 g/L/D
3.1	Re-design the sorbent to achieve the carbon capture capacity at $0.25 \text{ g CO}_2/\text{g Sorbent}$.	Q10(03/31/2024)	Completed The new sorbent reached $0.7g \text{ CO}_2$ /g sorbent adsorption capacity, and reached the milestone.

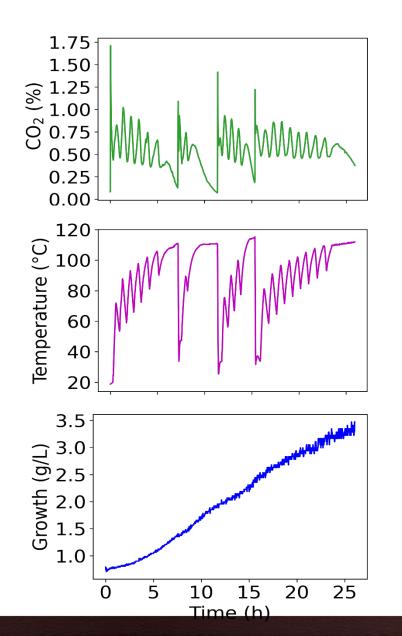








Evolution of The Integration System – Version 5_2L



□ Trigger threshold set to 0.8%

□ Growth over 16h: median at 1.84 g/L and peak at 2.00 g/L, surpassing the Q10 milestone (1.8 g/L)

To be tested/optimized:

Optimize the with CO₂ controlling system with capacity to maintain CO₂ at higher concentration

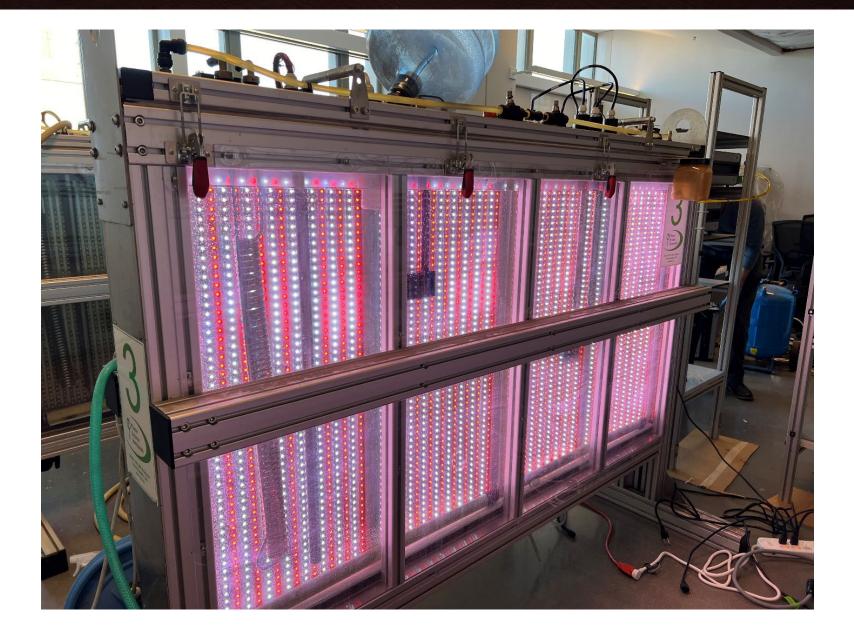






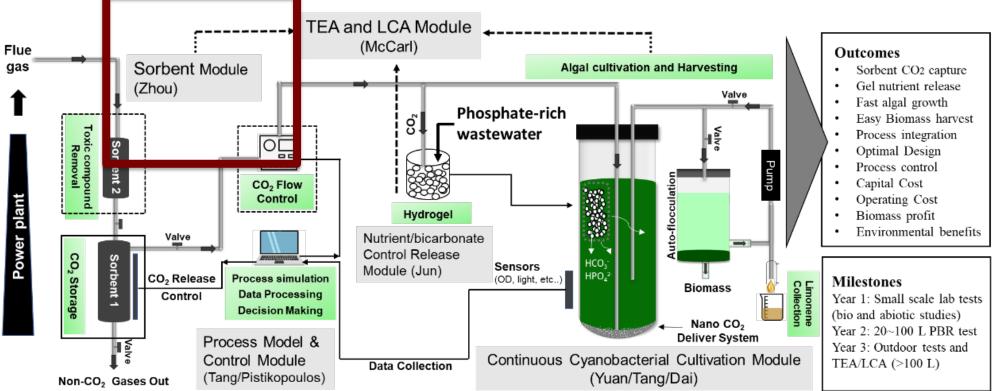






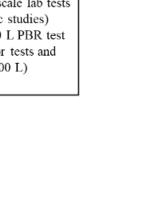


100 Liter PBR



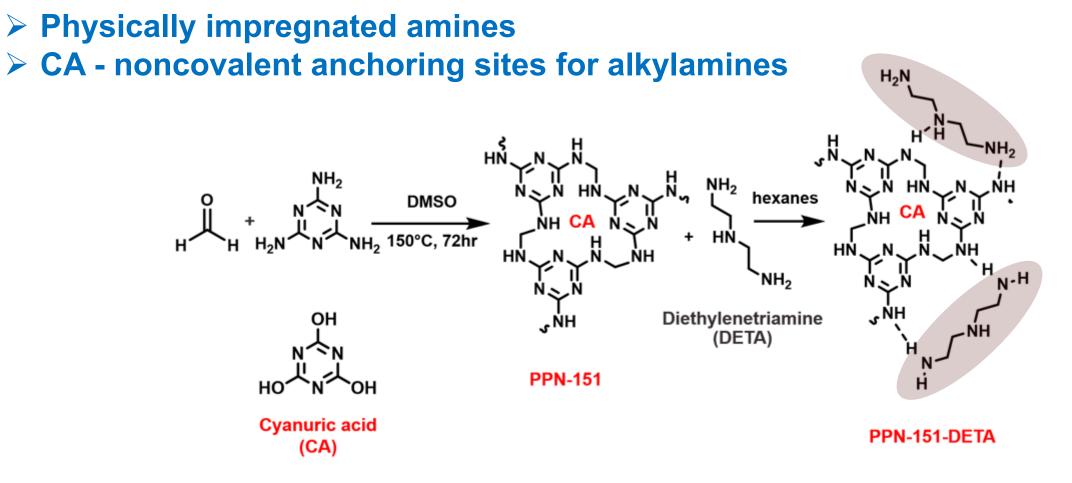
Zhou's group@TAMU







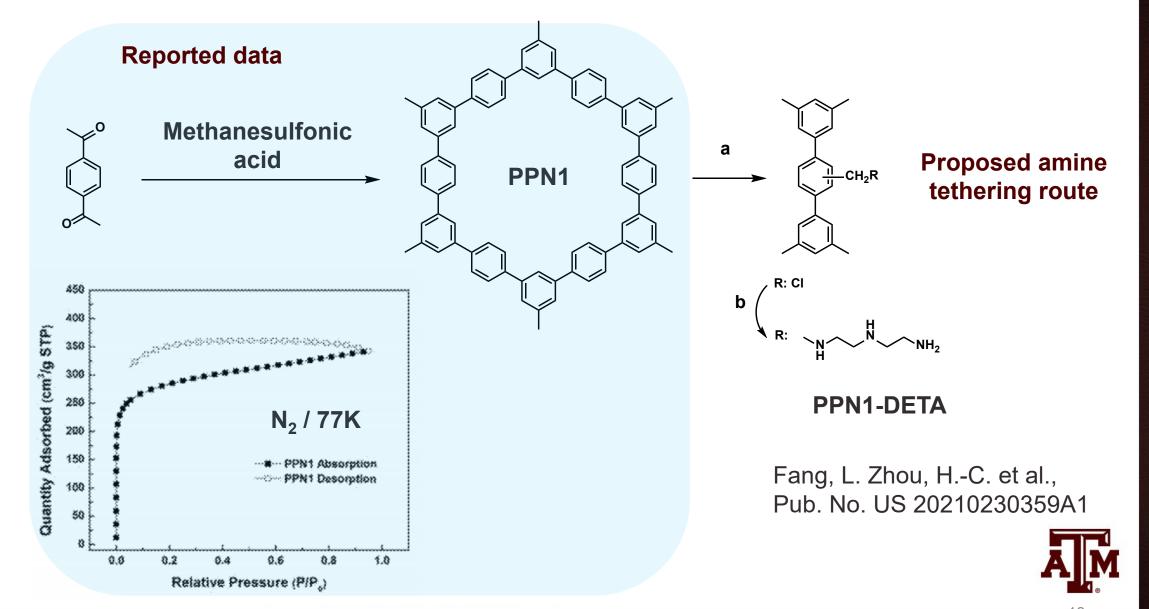
Physically Attached Amines in PPN-151-DETA



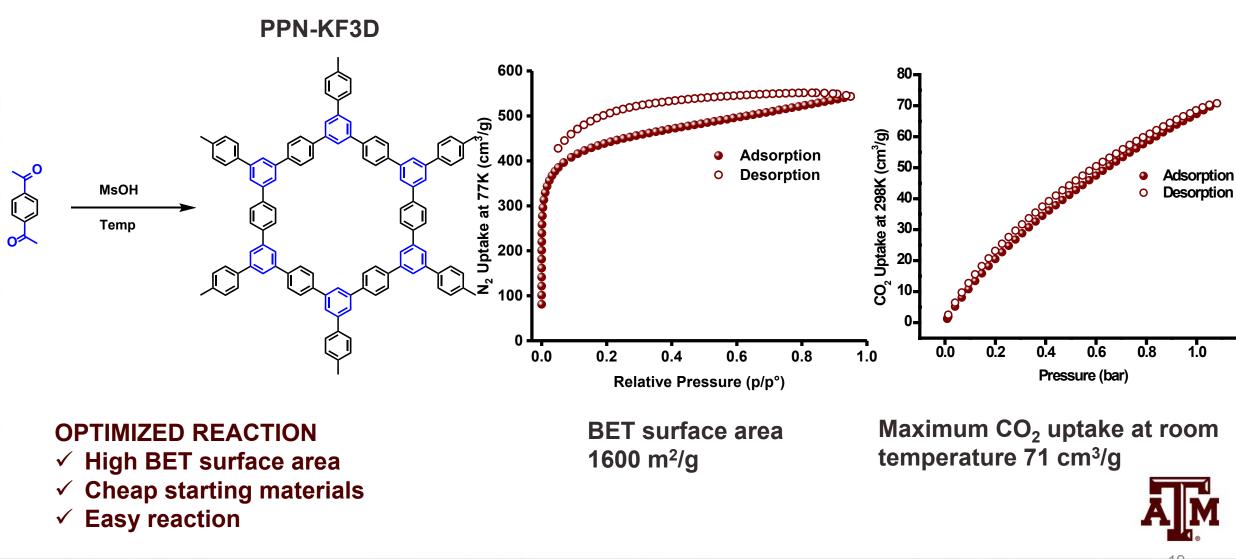
Advantage: Easy recycle of PPN-151 backbone; Disadvantage: Potential amine loss during application.

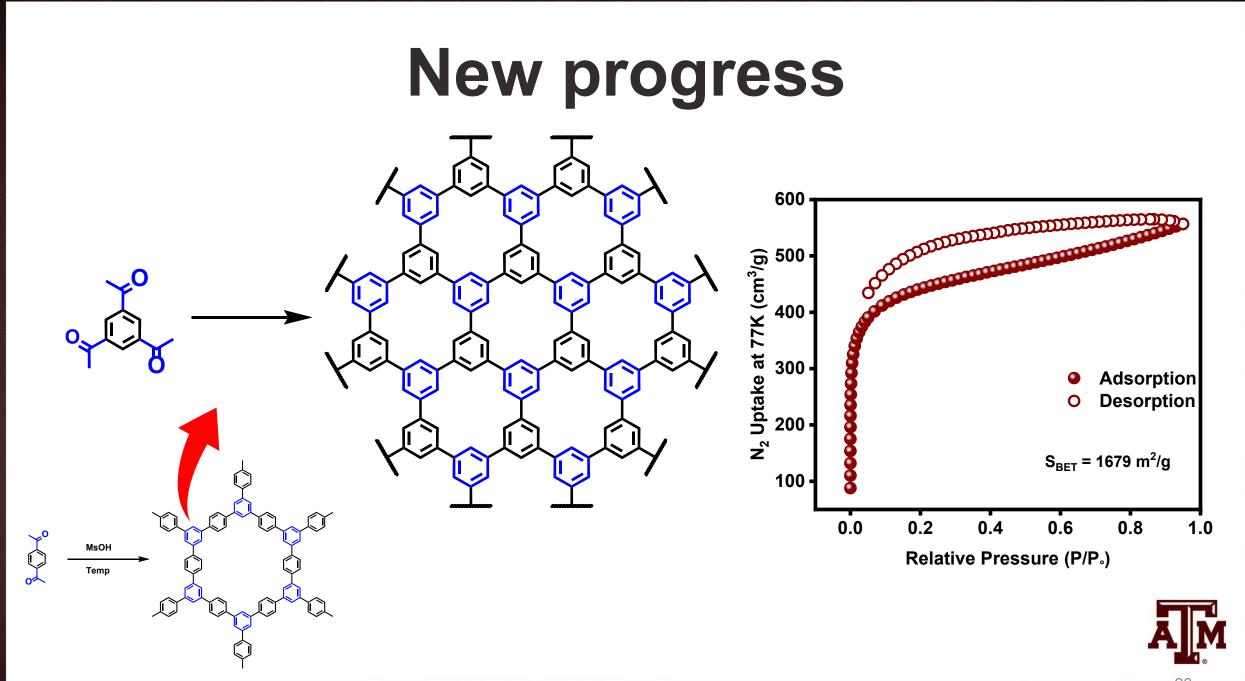


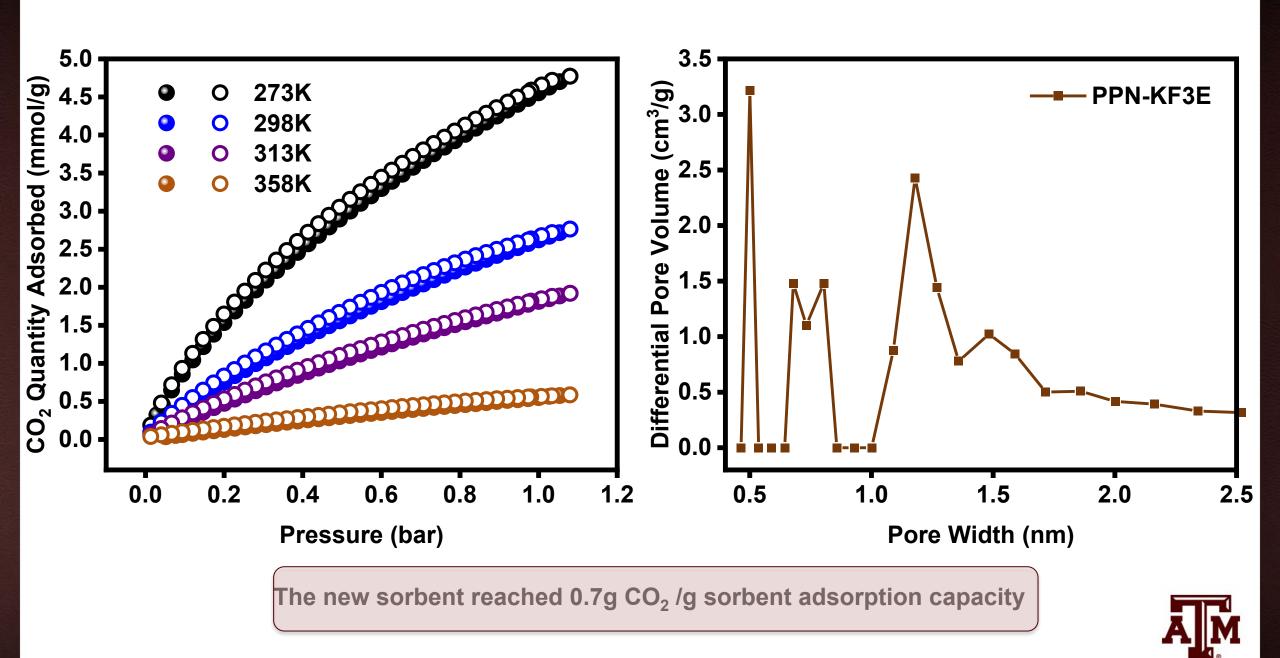
All Carbon PPN Scaffold

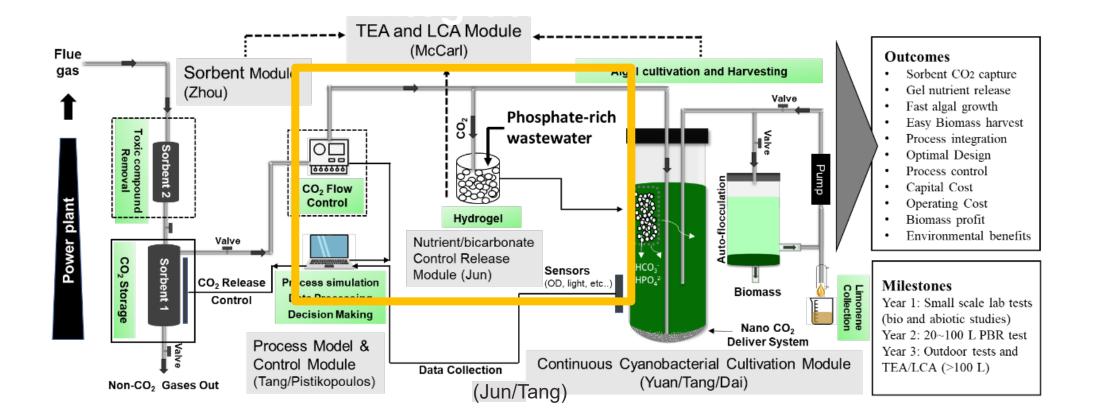


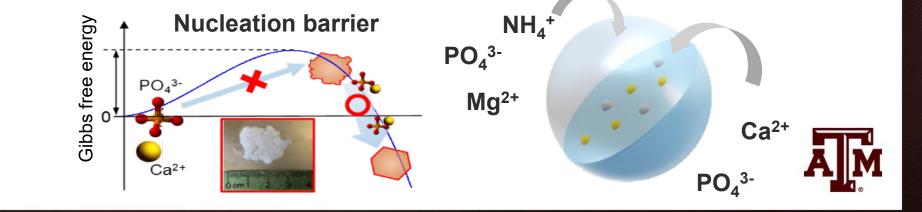
Porosity Optimization







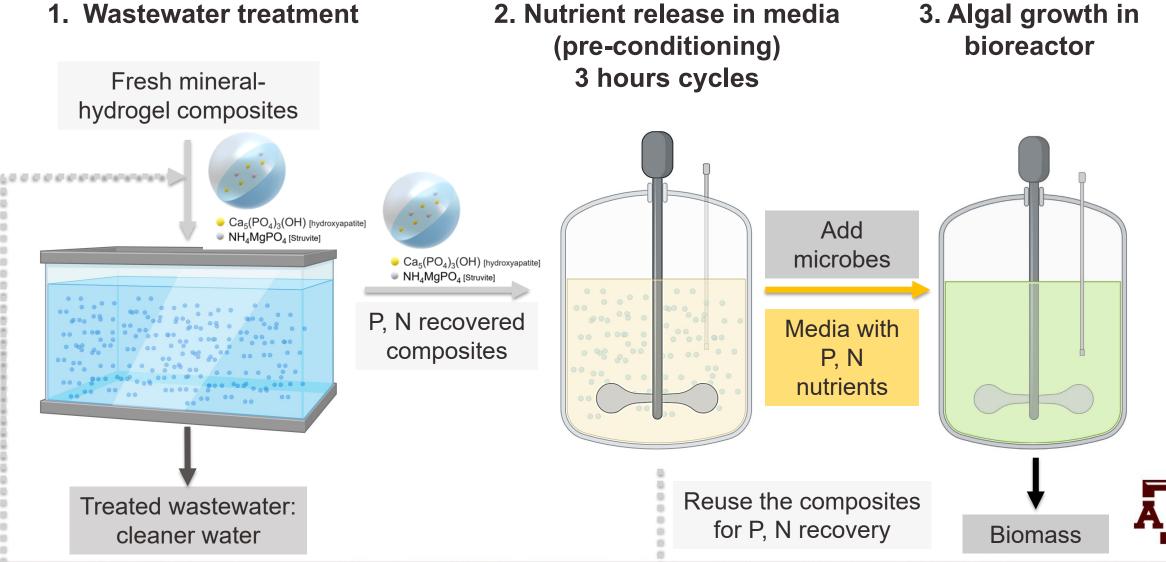




Jun's group @WashU

Nutrient delivery system from mineral-hydrogel composites to bioreactor

Created with BioRender.com



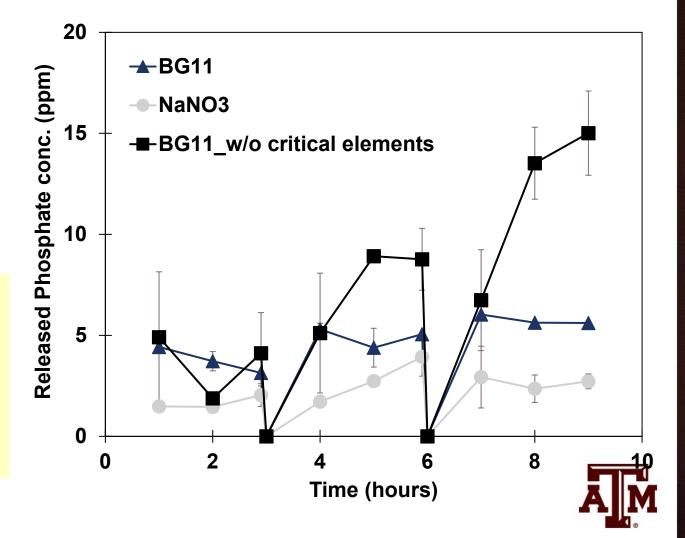
Release rates of mineral-hydrogel composites Ca-Alg/CaP+Wollastonite

Media

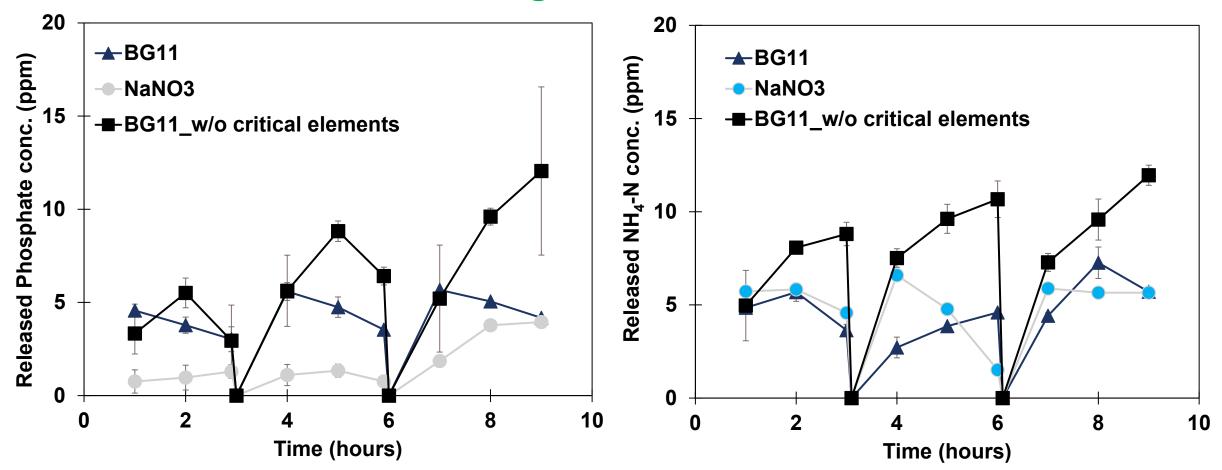
- 1. BG11 Media
- 22 mM NaNO₃ (same ionic strength as BG11)
- BG11 Media without critical elements (i.e., Ca, Mg, NH₄, PO₄)
 Media was replaced by every 3 hours
- Dose 10 % (v of hydrogel precursor/v of media)

In each cycle, the composites released around 5.0 ppm, 8.9 ppm, 15.1 ppm of P from Ca-Alg/CaP+Wollastonite

P release in each cycle is equivalent to P in BG11 media (5 ppm) or significantly over it.



Release rates of mineral-hydrogel composites Ca-Alg/CaP+Struvite

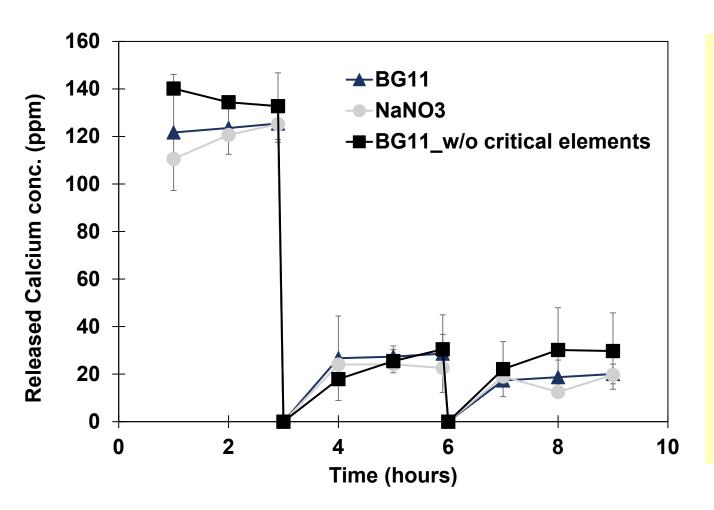


In each cycle, the composites released around 5 ppm, 9 ppm, and 13 ppm of P and 8.8 ppm, 10.7 ppm, and 12.0 ppm of NH_4 -N.

P release in each cycle is equivalent to P (5 ppm) in BG11 media or higher.

We can utilize Ca-Alg/CaP+Struvite with sufficient nitrogen delivery for algal species preferring ammonium sources

Release rates of mineral-hydrogel composites Ca-Alg/CaCO₃



- Calcium carbonate dissolution kinetic is relatively faster than struvite and calcium phosphate.
- Thus, the first cycle with 3 hours released the highest calcium and gradually decreased.
- Total Ca released amount is around 200 ppm in BG 11 without critical elements. It is equivalent to 305 ppm of bicarbonate release.

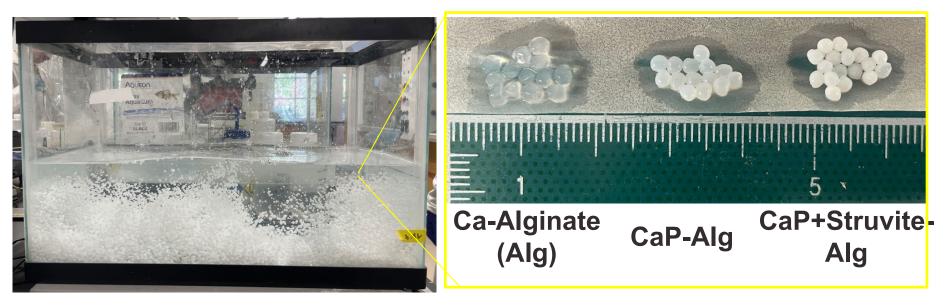


Fabrication of mineral-hydrogel composites at 250 g scale

Task 2.2 Hydrogel advancement and cultivation integration

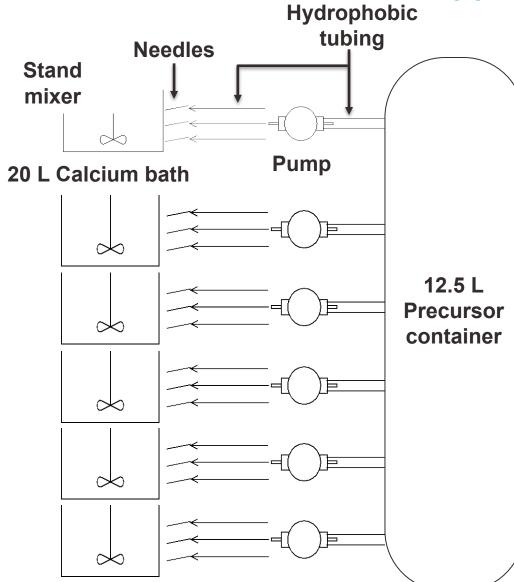
Achieve average dry weight percent of carbonate/ P/ N-containing minerals -- mineral 40 wt.% at 50g scale by 6/30/2023→ Achieved
 Milestone
 Achieve average dry weight percent of carbonate/ P/ N-containing minerals -- mineral 50 wt.% at 250g scale by 9/30/2024→ In Progress (80% completion)

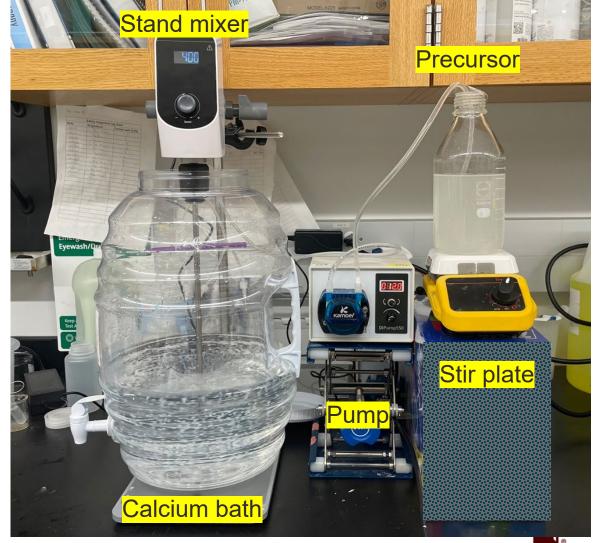
Mineral-hydrogel composite synthesis





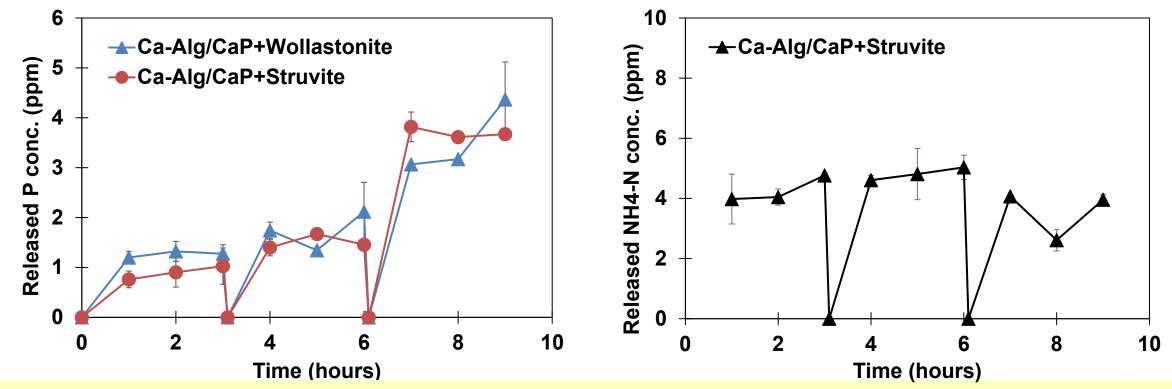
Fabrication of mineral-hydrogel composites at 250 g scale





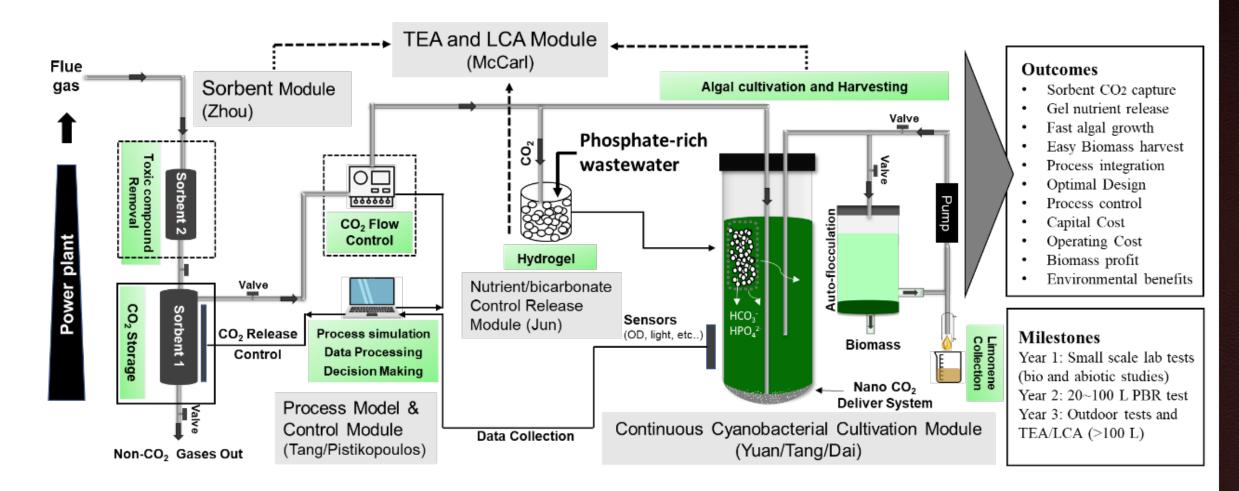
Release rates of mineral-hydrogel composites : Large scale dissolution

- Media BG11 Media without critical elements (Ca, Mg, NH₄, PO₄)
- Media was replaced by every 3 hours.
- Dose 10 % (v of hydrogel precursor/v of media)
- Total volume of media: 7.5 L (150 times increase than bench scale)



• The P and N released concentrations lower than those at the bench scale. Thus, they required more extended time dissolution to reach a similar level of P and N in BG11.

• It can be improved by changing the hydrogel fixation system to enhance the nutrient transformation.



Modeling



Microbe Action and Growth Modeling Application (MAGMA)

Built with MATLAB

Simulation of bioreactors represented by ordinary differential equation (ODE) systems

Nonlinear regression for fitting models to data



Home	Components	Environment	Mass Transfer	Reactions	Solver	Model Re	eview	Parameters	Regression	Plot Customization	
	Co	omponent S2973	(Liquid Phase)							•	
Governin	g Function:										
				$\frac{dX_1}{dt} = \mu_{max} X_1$	(I	$-k \cdot \mathbf{Y}_{i}$					Saved
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Edit Euno	tions							System Variabl	•	Expression	
mu_max	*X_1*(l/(K_l+l))-k_	_d*X_1				Upra	te			· · · ·	
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					*	Add LX	555 JH	N (Liquid Phase		C_2	_
Set Piecewise Limits on Functions								P (Liquid Phase) C_3			_
Expres	sion					▼ A	dd	C (Liquid Phase	:)	C_4	
	lf	<= v t	▼ <= ▼	Els	se 0	R ar	nove	Equilibrium Hyd	rionium Concen.	. H3O_eq	
						Outside Bounds		Equilibrium Hyd	roxide Concentr.	OH_eq	
Lower Bo	bund	Upper Bound	Variat	ble	Value O	utside Bou	nds	pН		рН	
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Helper Fu	inctions							JI 			
Add N				▼ Update	Por	nove					
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		:	=								
Clear Console Show Console							Stop Calculation				

Fitting algal limonene model to experimental data

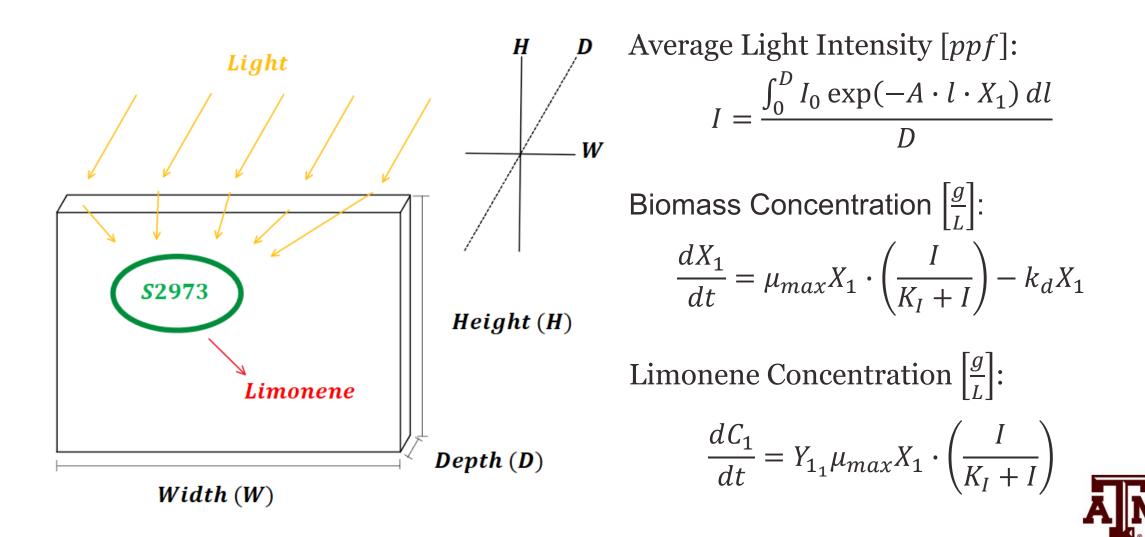
Focus on effect of light availability on algal biomass growth

Experimental data obtained from flat-panel PBR trials

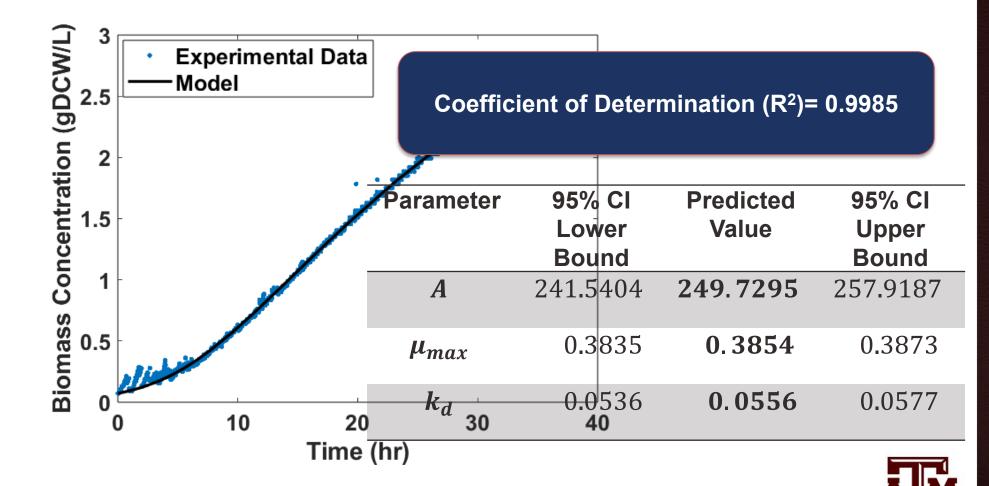
Assumes light is the only significantly growthlimiting substrate in experimental trials



A simplified model is fit to experimental data to obtain kinetic parameters describing the impact of light intensity on biomass growth



Model fitting was successful, yielding 3 important kinetic parameters



Further experiments can determine more kinetic parameters describing the effect of other relevant factors on biomass growth

Batch cultivation with single limiting substrate

- Impact of C, N, P nutrients on biomass growth rate
- Nutrient-to-biomass yields

Batch cultivation with light as limiting substrate

• Obtain accurate value for K_I parameter (assumed in current model)

Goal: develop model useful for photobioreactor control and bioprocess optimization

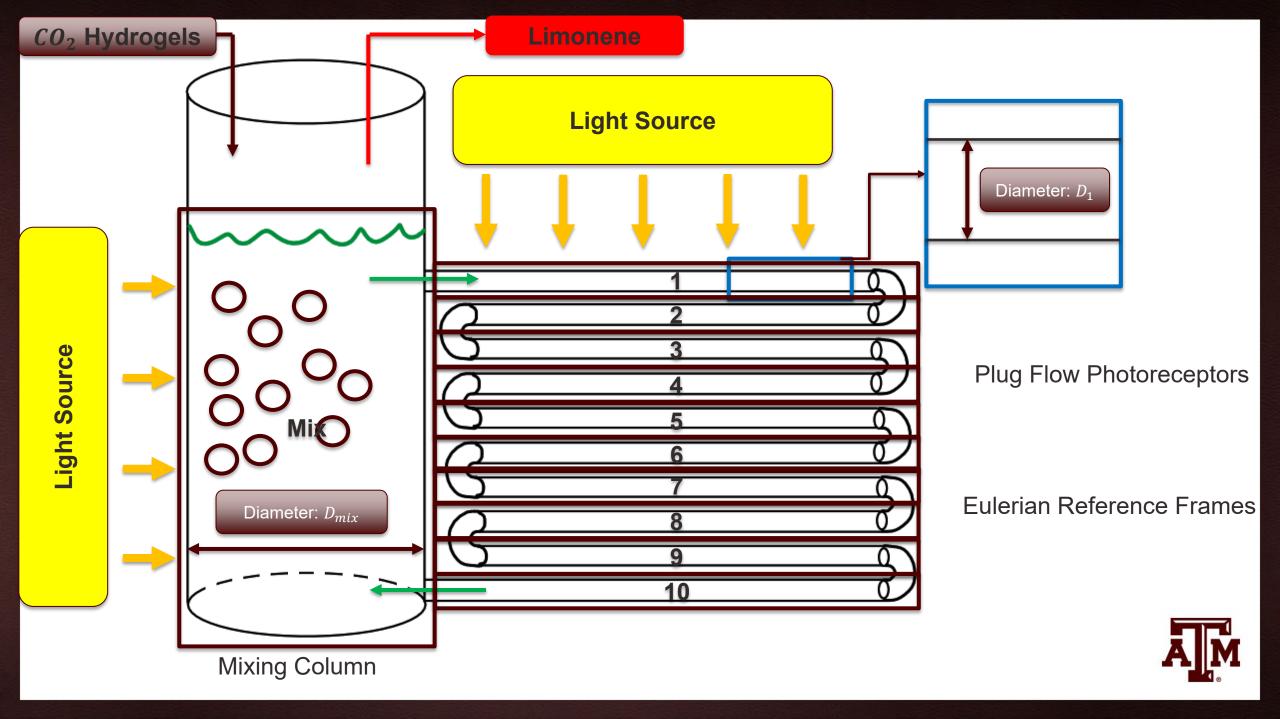


Predicting performance of photobioreactor for producing Limonene from light and CO₂ via cyanobacteria

Simulating impact of light and nutrient concentration on biomass and product formation

- Biomass light shading
- Effect of PBR geometry on light availability
- Experiment-calibrated light modeling





Biochemical Reaction Functions for Algal PBR System for n = mix, 1, 2, ..., 10:

$$Biomass: \frac{dX_n}{dt} = \mu_{max} X_n \left(\frac{N_n}{N_n + K_N}\right) \left(\frac{P_n}{P_n + K_P}\right) \left(\frac{CO_{2,n}}{CO_{2,n} + K_{CO_2}}\right) \left(\frac{I_n}{I_n + K_I}\right) - k_d X_n$$

$$CO_2: \frac{dCO_{2,m}}{dt} = -Y_{CO_2} \mu_{max} X_n \left(\frac{N_n}{N_n + K_N}\right) \left(\frac{P_n}{P_n + K_P}\right) \left(\frac{CO_{2,n}}{CO_{2,n} + K_{CO_2}}\right) \left(\frac{I_n}{I_n + K_I}\right)$$

$$\frac{dCO_{2,mix}}{dt} = \mathbf{0}$$

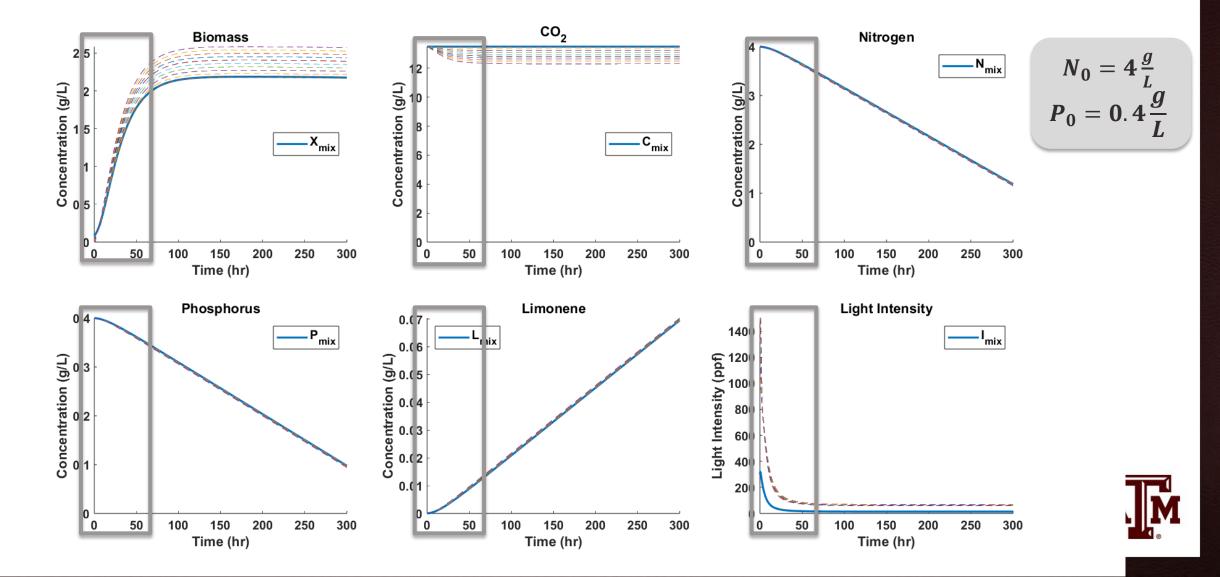
$$Nitrogen: \frac{dN_n}{dt} = -Y_N \mu_{max} X_n \left(\frac{N_n}{N_n + K_N}\right) \left(\frac{P_n}{P_n + K_P}\right) \left(\frac{CO_{2,n}}{CO_{2,n} + K_{CO_2}}\right) \left(\frac{I_n}{I_n + K_I}\right)$$

$$Phosphorus: \frac{dP_n}{dt} = -Y_P \mu_{max} X_n \left(\frac{N_n}{N_n + K_N}\right) \left(\frac{P_n}{P_n + K_P}\right) \left(\frac{CO_{2,n}}{CO_{2,n} + K_{CO_2}}\right) \left(\frac{I_n}{I_n + K_I}\right)$$

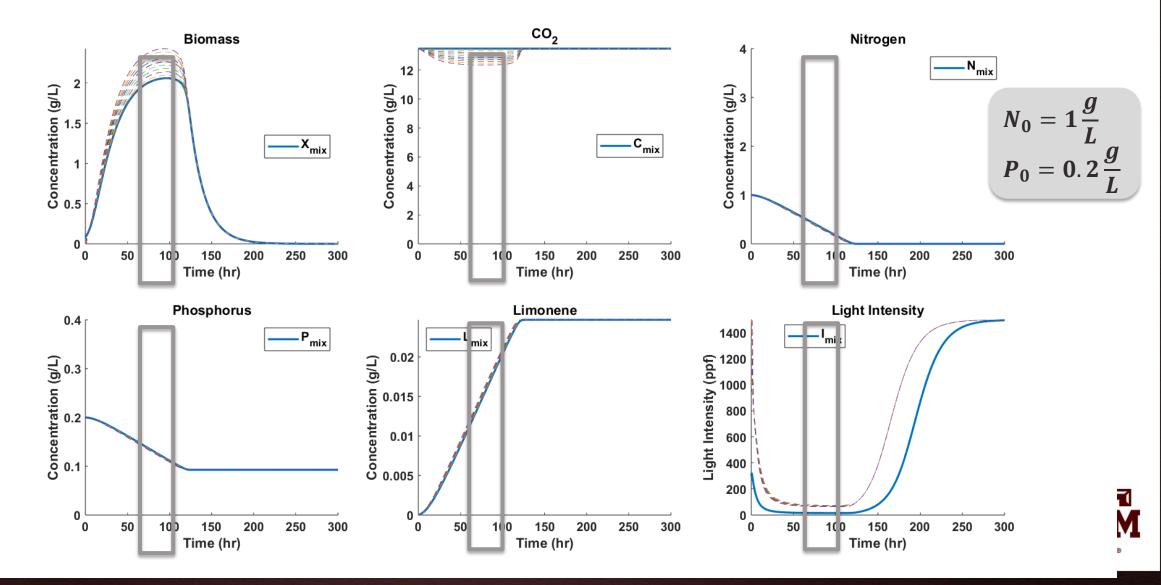
$$Light: I_n = \frac{\int_0^{D_n} \int_0^{\sqrt{D_n^2 - w^2}} I_0 \exp(-A \cdot (w \cdot X_n)) dh dw}{\pi \left(\frac{D_n}{2}\right)^2}$$

$$Calculates the average light intensity in each Eulerian reference frame$$

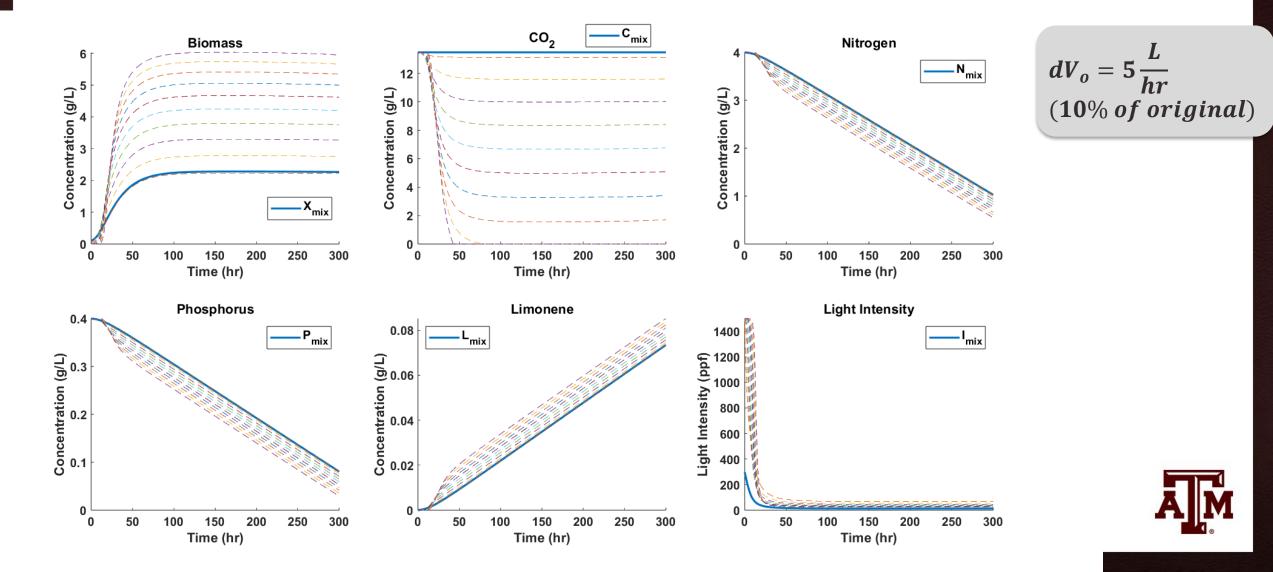
Growth Limited By Light Resources



Growth Limited By Nutrient Resources



Slower PFR Flowrate (Nutrient Limitation Case)



Lessons Learned

- Scale up: what you can plan and what you cannot
 - 1. Cultivation scale-up
 - 2. Sorbent synthesis
 - 3. Hydrogel synthesis
- Modeling can be very helpful for scale up
- CO2 concentration



Future Plans

• Scale-up: 20-liter and 100-liter cultivation/sorbent and hydrogel synthesis and testing

• Work with NCCC for on-site CO2 conversion

• System integration



Our Team



TAMU Plant Pathology & Microbiology Microbial engineering and development of continuous algal cultivation platform



TAMU AgricultureEconomicsLife cycle analysis andenvironmental analysis



Zhou

TAMU Chemistry Amine-based porous sorbent advancement WUSTL Chemical, Energy & Environmental Engineering Unique hydrogel technologies and process design

Yuan

NCCC at Southern

Company

Scale up and on-site

testing

Jun

Morten



Tang

Kumfer



Pistikopoulos

Chemical Engineering System modeling and TEA

poulos

Acknowledgement

- DOE Office of Fossil Energy
- Dr. Lei Hong

• Questions?

