

Opportunities and Challenges Related to Anaerobic Digestion of Food Waste

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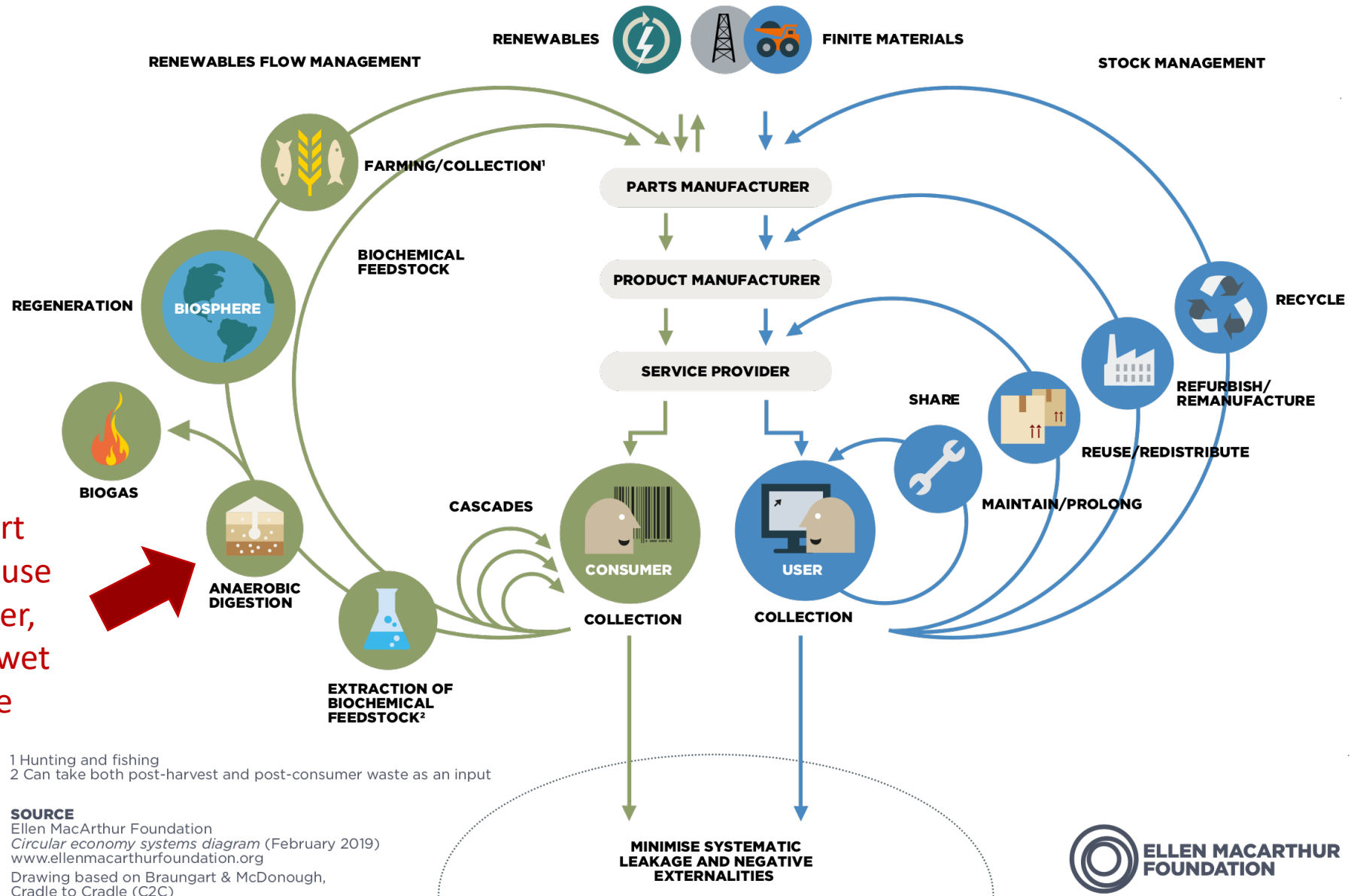
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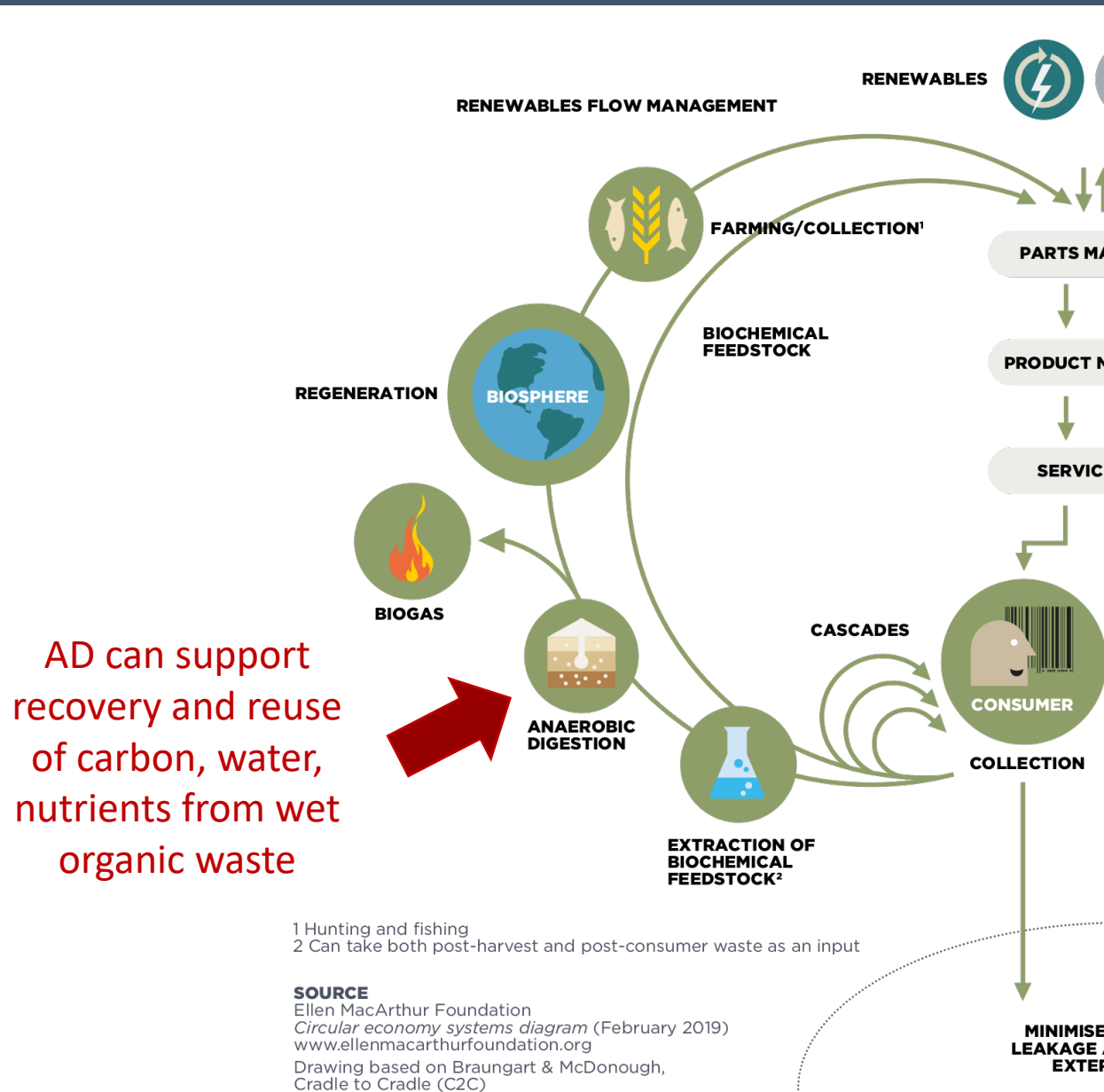
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Anaerobic Digestion (AD) is recognized as beneficial process within the Circular Bioeconomy

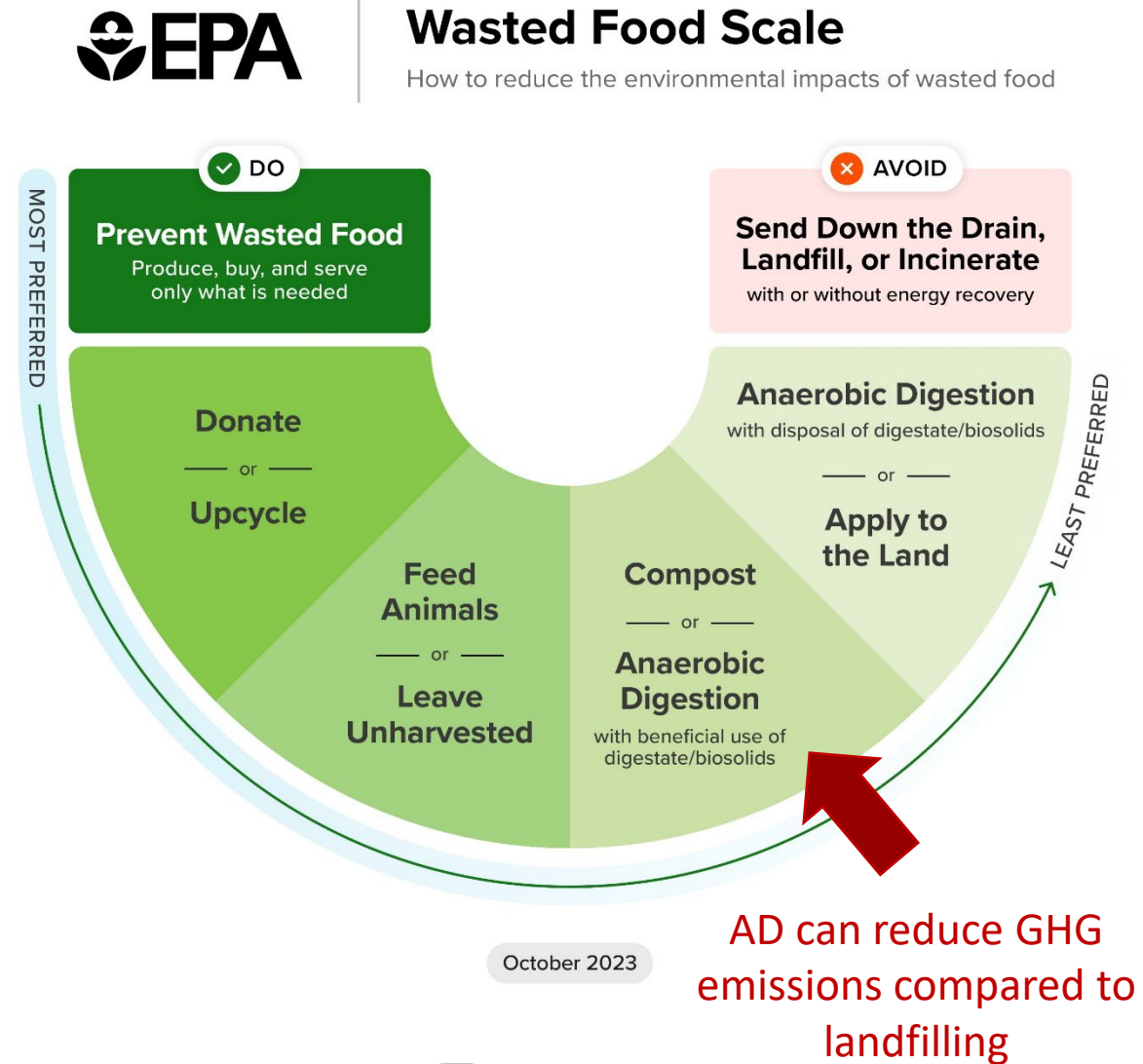
AD can support recovery and reuse of carbon, water, nutrients from wet organic waste



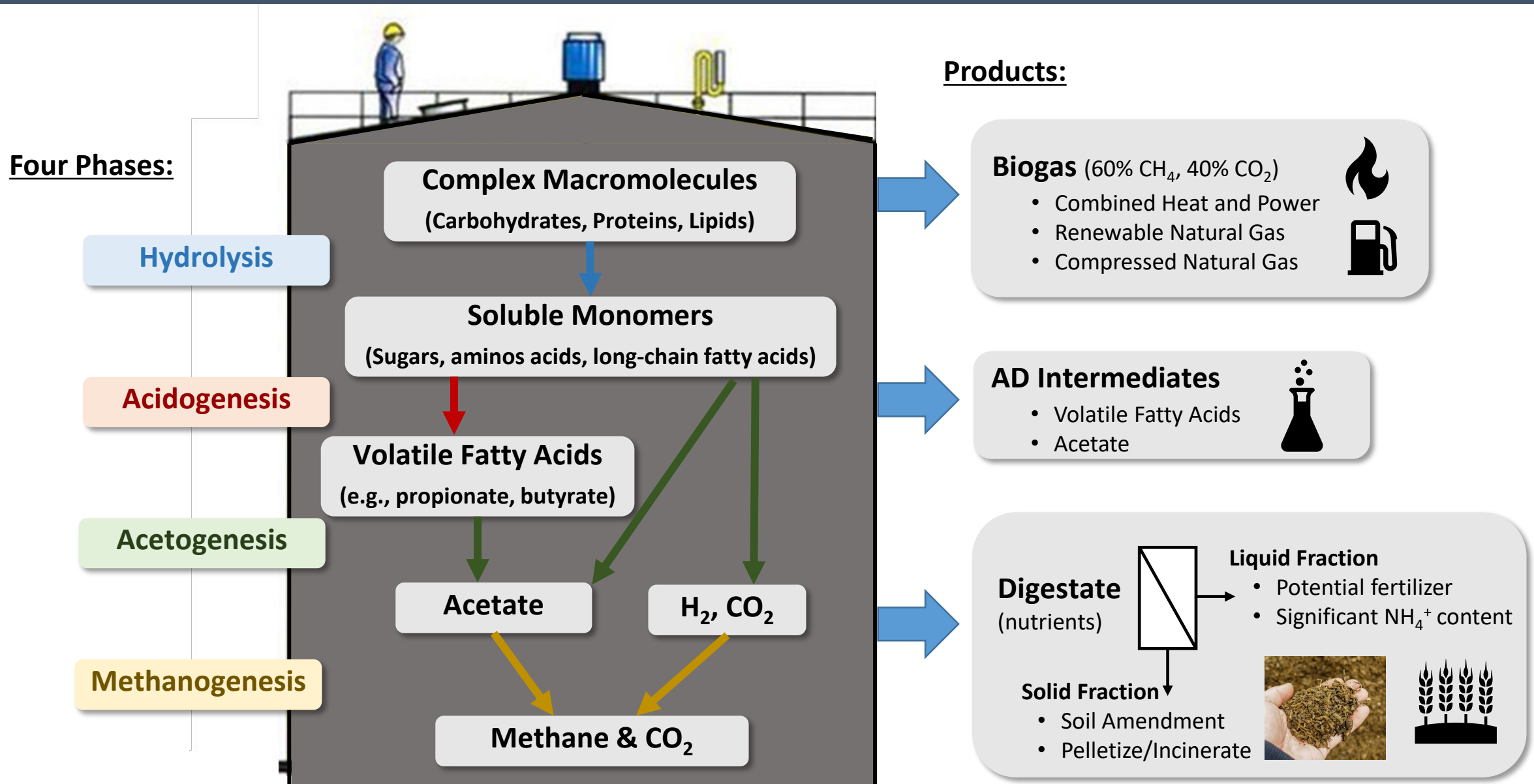
Anaerobic Digestion (AD) is recognized as beneficial process within the Circular Bioeconomy



AD can support recovery and reuse of carbon, water, nutrients from wet organic waste



Anaerobic Digestion is a balance of microbes working together to convert organics to biogas



Examples of Commercial Scale Digesters



**Saerbeck Bioenergy Park,
Germany**



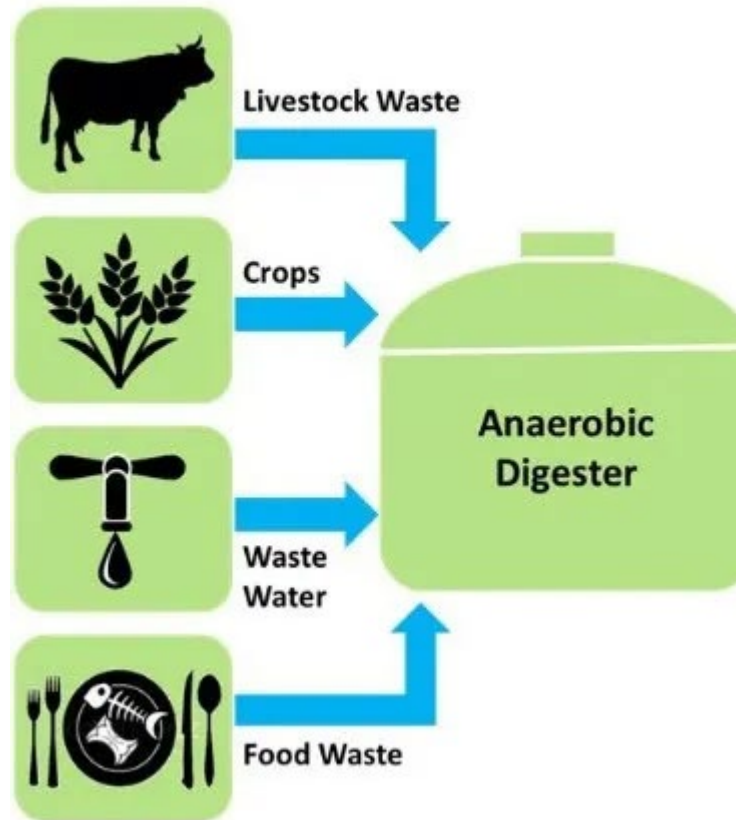
**Alliance Dairy
Trenton, Florida**



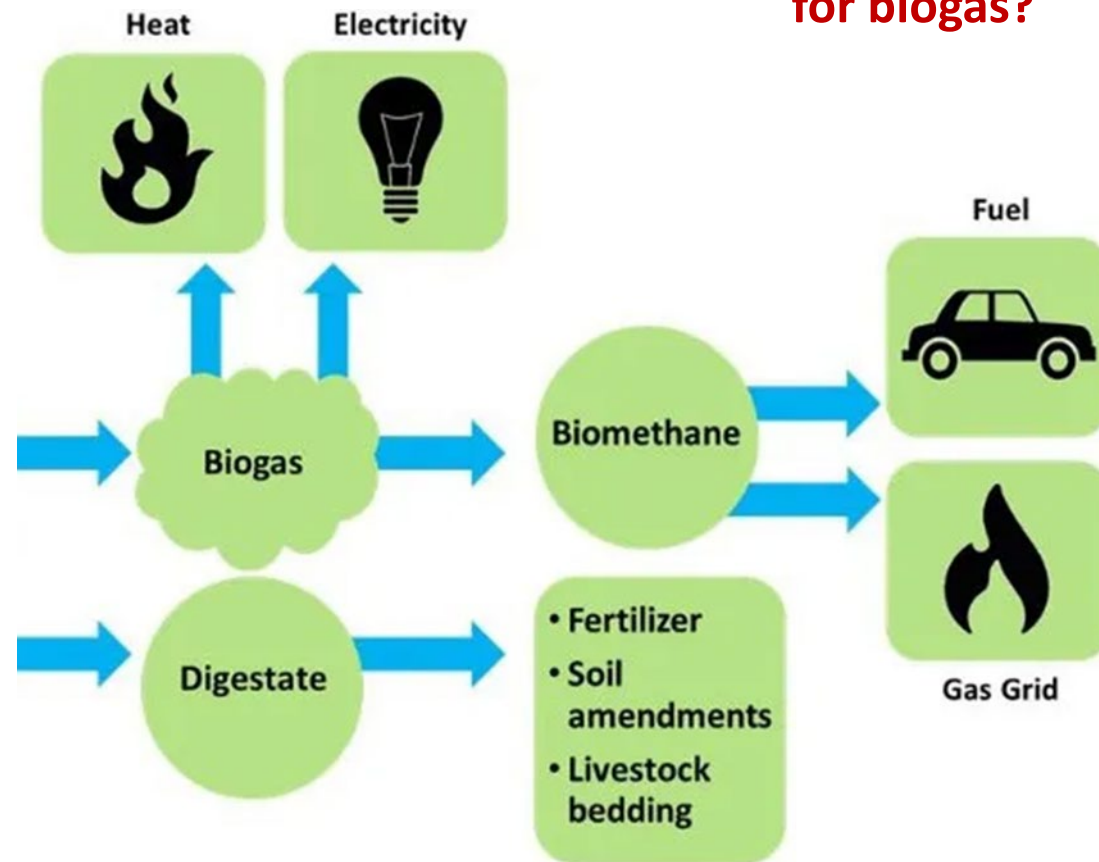
**Bigadan
Copenhagen, Denmark**

The Three Questions...

(1) What will the feedstock be and how much?

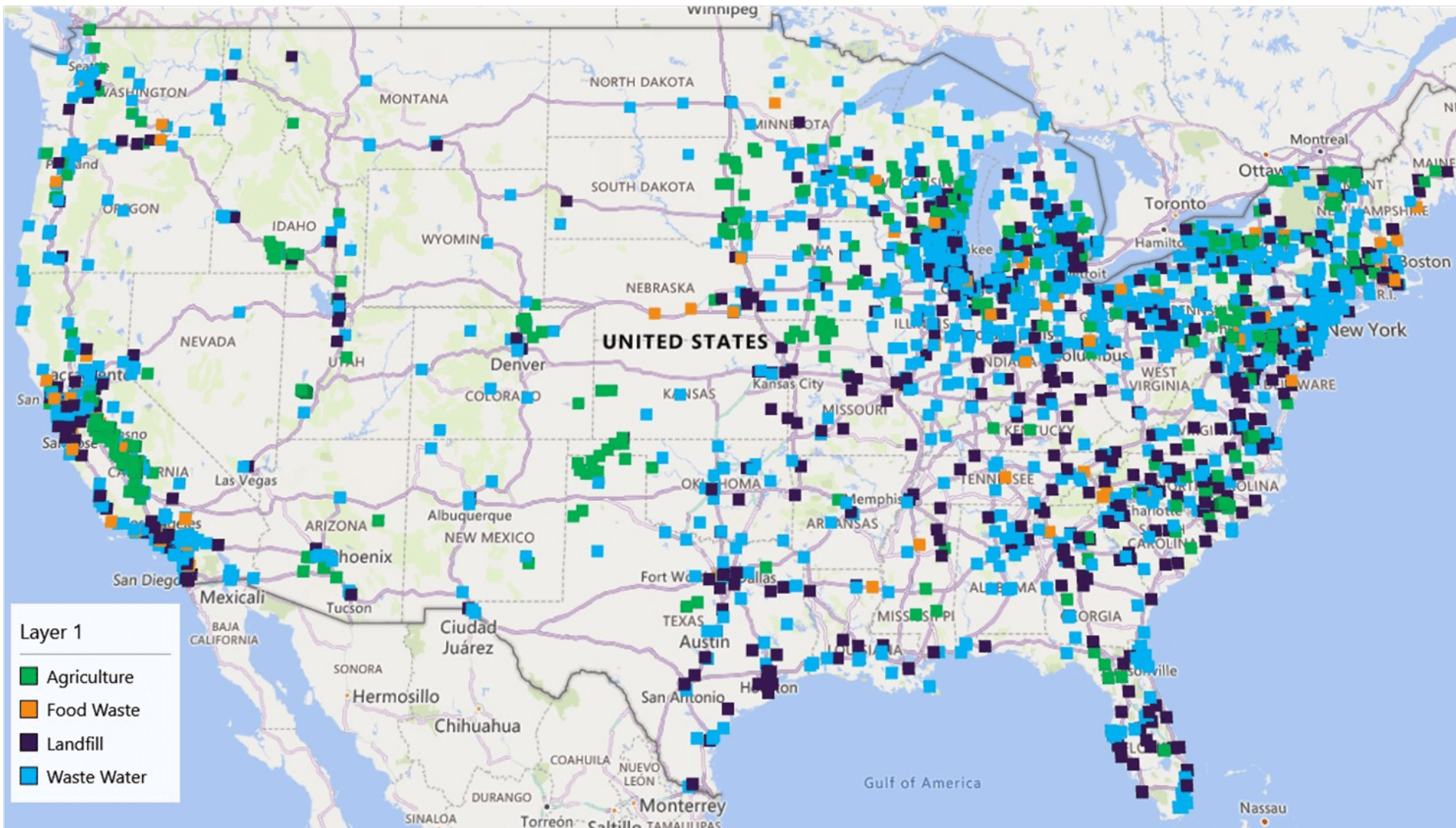


(2) What is the market/value for biogas?



(3) What is market/management options for digestate?

Anaerobic Digestion (AD) in the U.S.



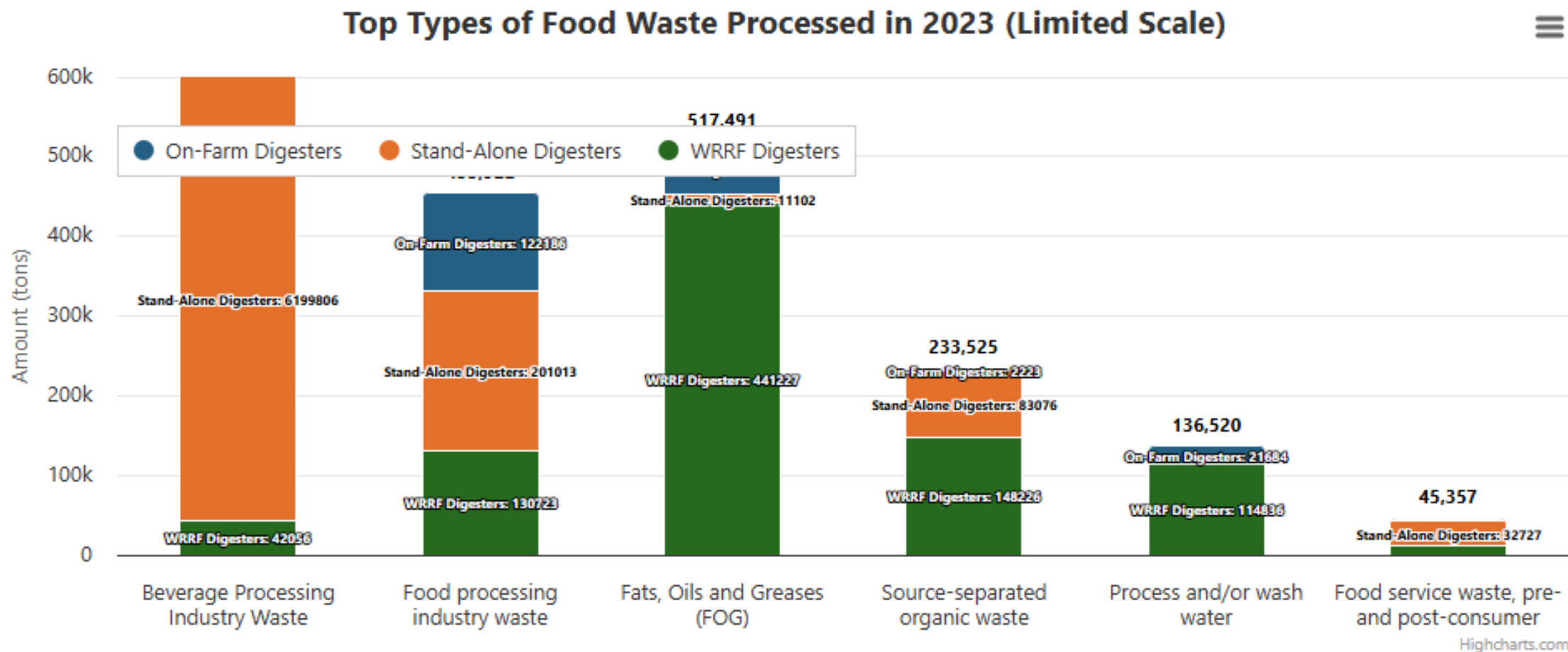
According to American Biogas Council:

2,500 AD sites in U.S.

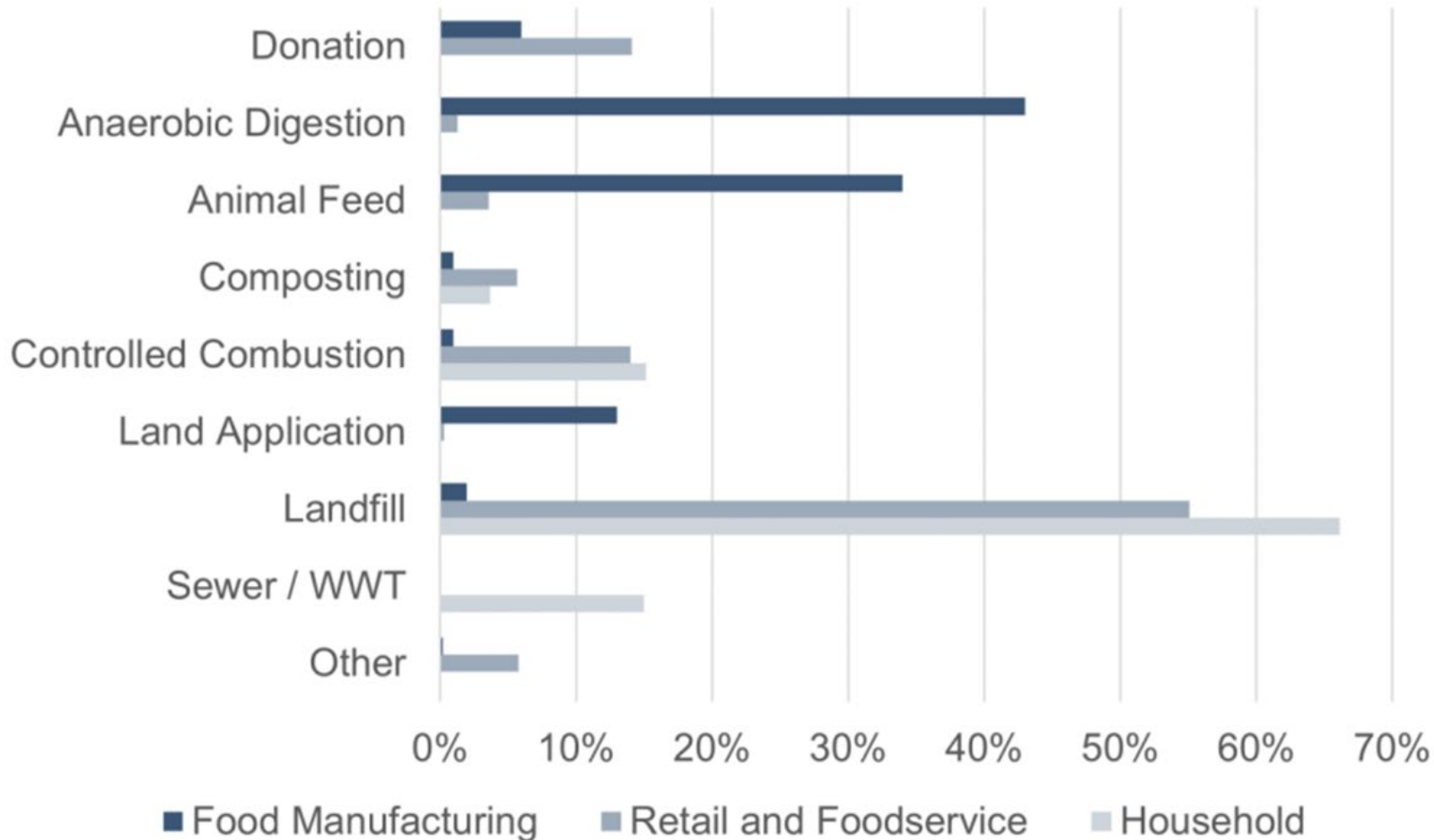
- 609 on farms
- >1,180 at WRRFs
- 113 food waste
- 583 landfill gas

Germany has nearly 10,000 operating digesters

Currently Anaerobic Digestion of post-consumer food waste is limited



Most post-consumer food waste is sent to landfills due to AD challenges



- Challenges related to AD:**
- (1) Food Waste Variability**
 - (2) High Capital Costs**
 - (3) Digestate Management**



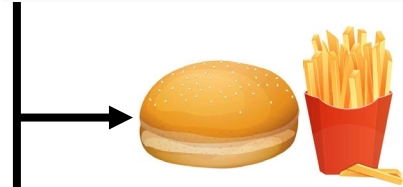
FIGURE 1-1. PERCENTAGE OF WASTED FOOD MANAGED BY EACH PATHWAY, BY SECTOR IN WHICH THE WASTED FOOD IS GENERATED

Challenge 1: Food Waste Variability will impact methane yield and stability



Carb:Protein:Lipid

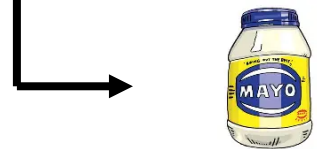
Baseline: 60:20:20



High Carb: 75:20:5

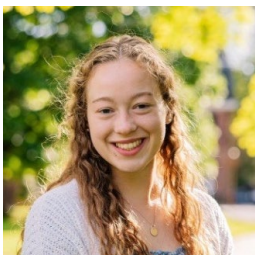
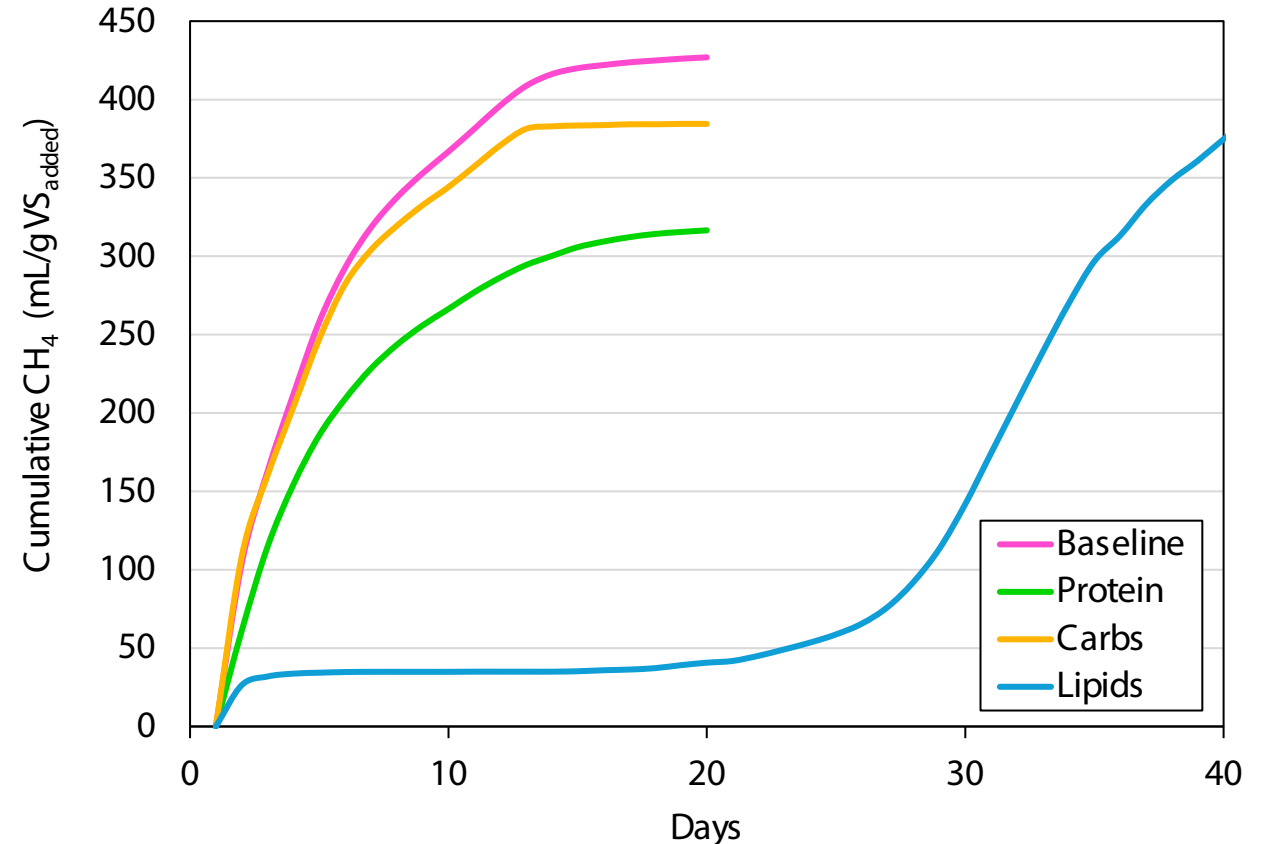


High Protein: 30:60:10



High Lipid: 40:5:65

Theoretical ranking: **Lipid > Carb > Baseline > Protein**



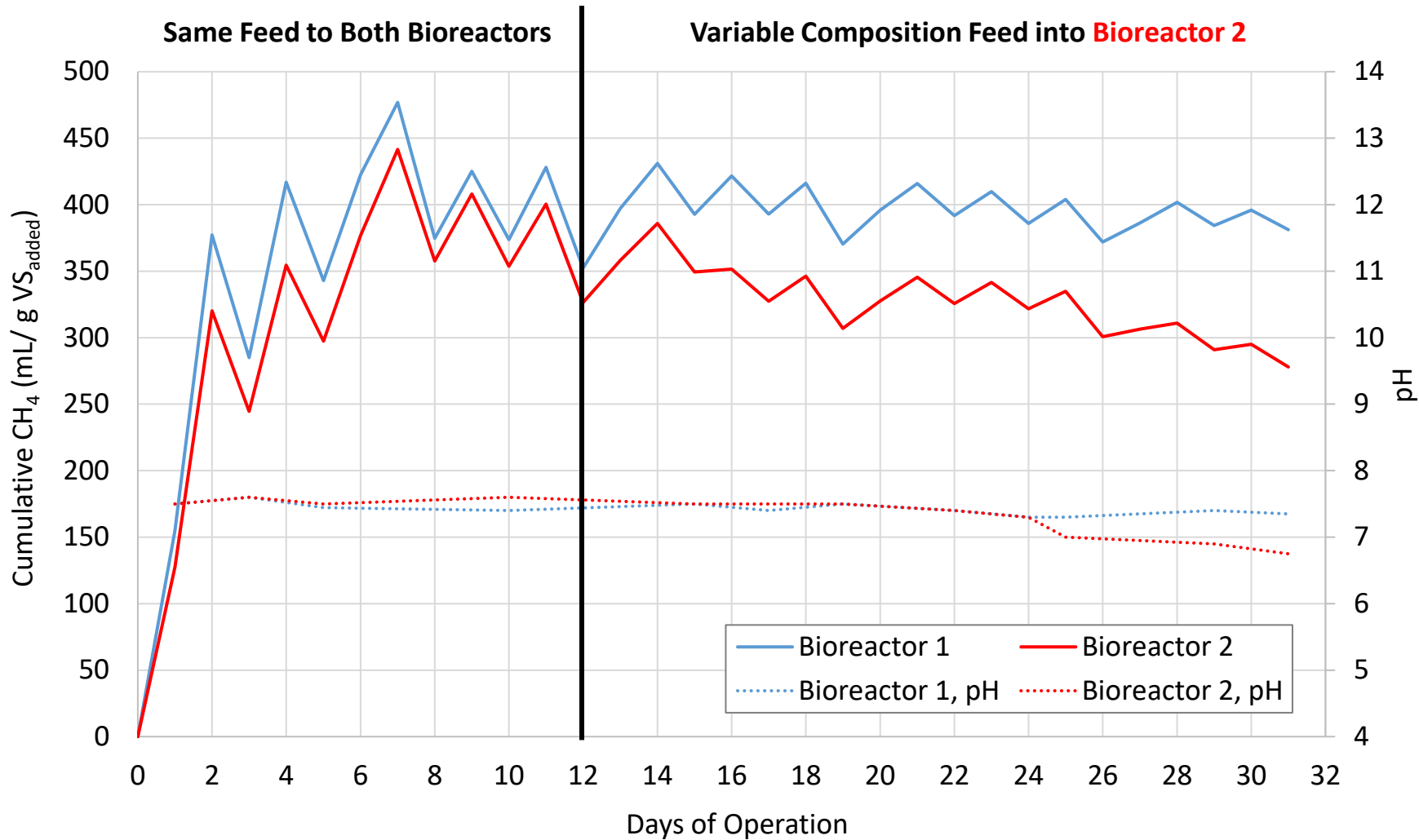
Chloe Hughes and
Elizabeth Perkins

Undergraduate Students

We found:

- **Baseline condition produced higher methane**
- **High lipid condition experienced extensive lag period**

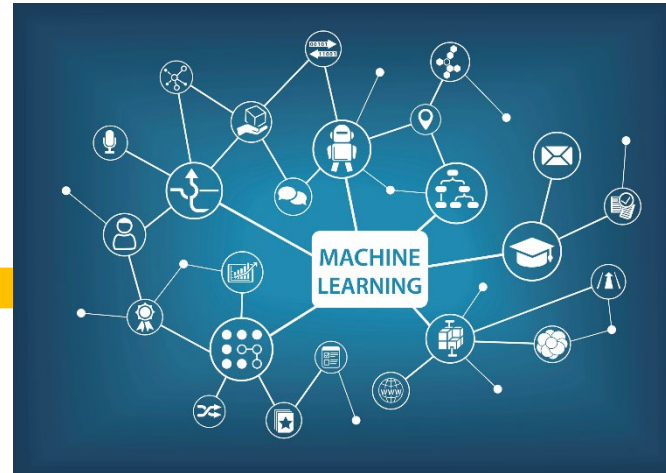
Challenge 1: Food Waste Variability will impact methane yield and stability



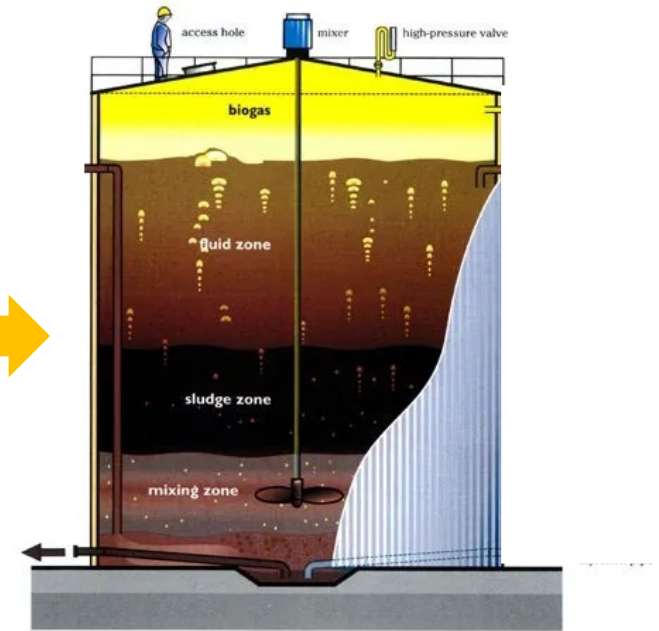
15 L working volume
Temp: 37°C
HRT: 30 days
OLR: 1 g VS/L/day

Prior Hinkley Project: Developing a Machine Learning Predictive Model for AD

Food Waste Composition



AD Process Outcomes



Dr. Nikolay Bliznyuk
Associate Professor



Dr. Jithran Ekanayake
Former PhD student



Camil Coss-Flores
Undergraduate

Machine Learning Model Development

1. Literature Review

Web of Science Database w/ key words:

“anaerobic digestion” AND “food waste” AND:
 “household” or “consumer” or “restaurant”

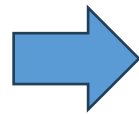
Round 1 = 100 records

Round 2 = 202 records, focused on feedstock macromolecules



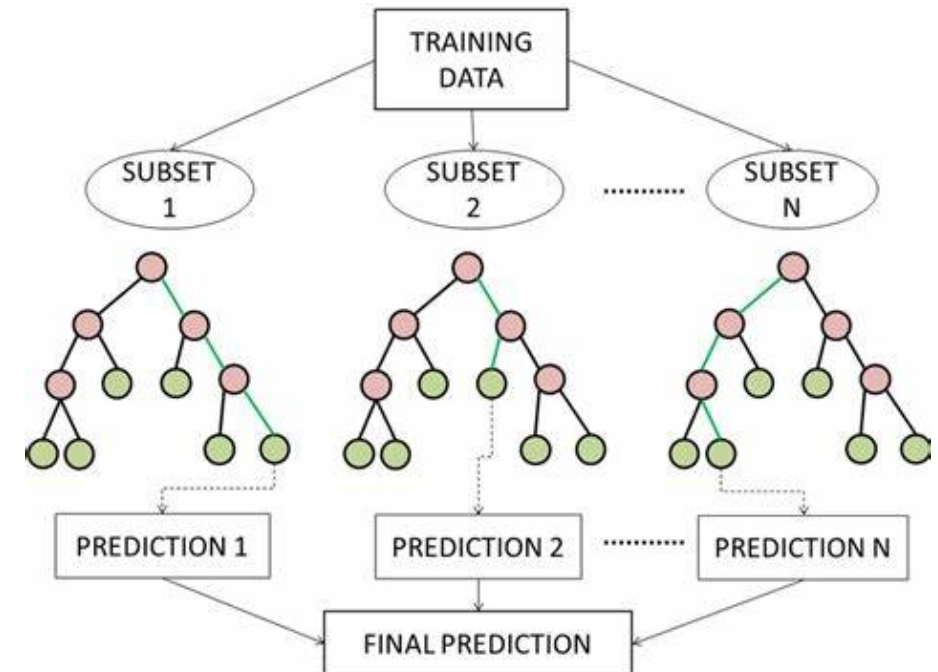
16 predictor variables identified

Variable Type	Categorical	Numerical
Feedstock Characteristics	- Feedstock Source	- Volatile Solids (VS) (% w/w) - Total Solids (TS) (% w/w) - VS/TS (%) - Carbohydrate Content (% TS) - Protein Content (% TS) - Lipid Content (% TS) - Feedstock C/N
Operating Conditions	- Reactor Type - Pretreatment Type	- Solid Retention Time (days) - Number of Process Stages - Digester Volume (L) - Hydraulic Retention Time (days) - Organic Loading Rate (g-VS/L-d) - Temperature (C)



2. Random Forest Regression

- Dataset was imputed and split into training (80%) and testing (20%) datasets using k-fold cross validation
- RandomForestRegressor method available in the Python Machine Learning Library, sci-kit learn, was used to build a regression tree with all predictors, without prescreening

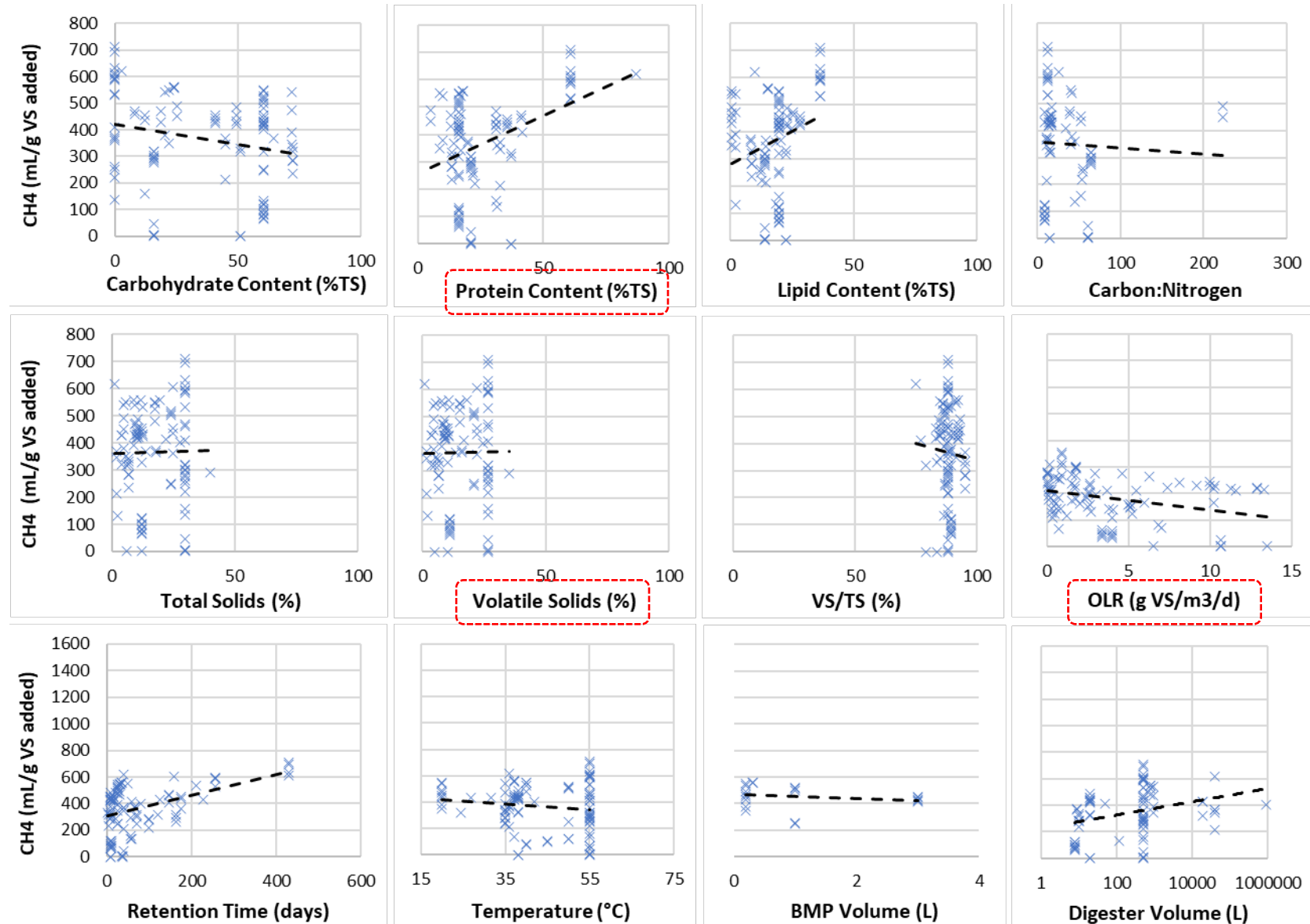
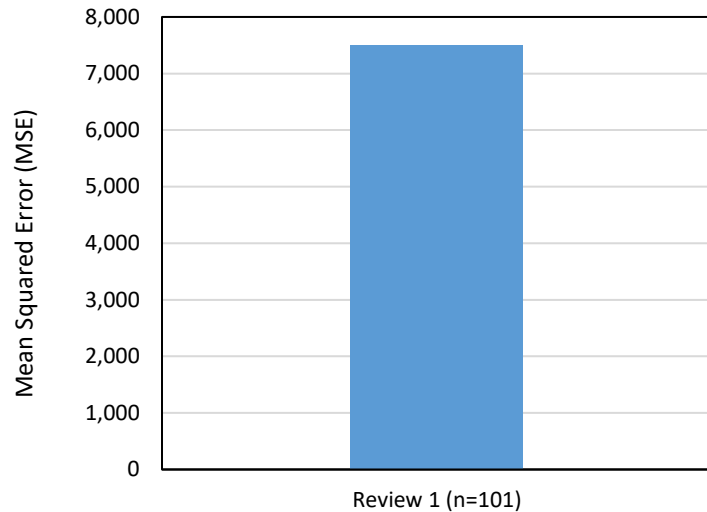


Random Forest using initial dataset (n=101) resulted in high mean squared error

Key predictor variables included:

- Protein Content
- Organic Loading Rate
- Volatile Solids

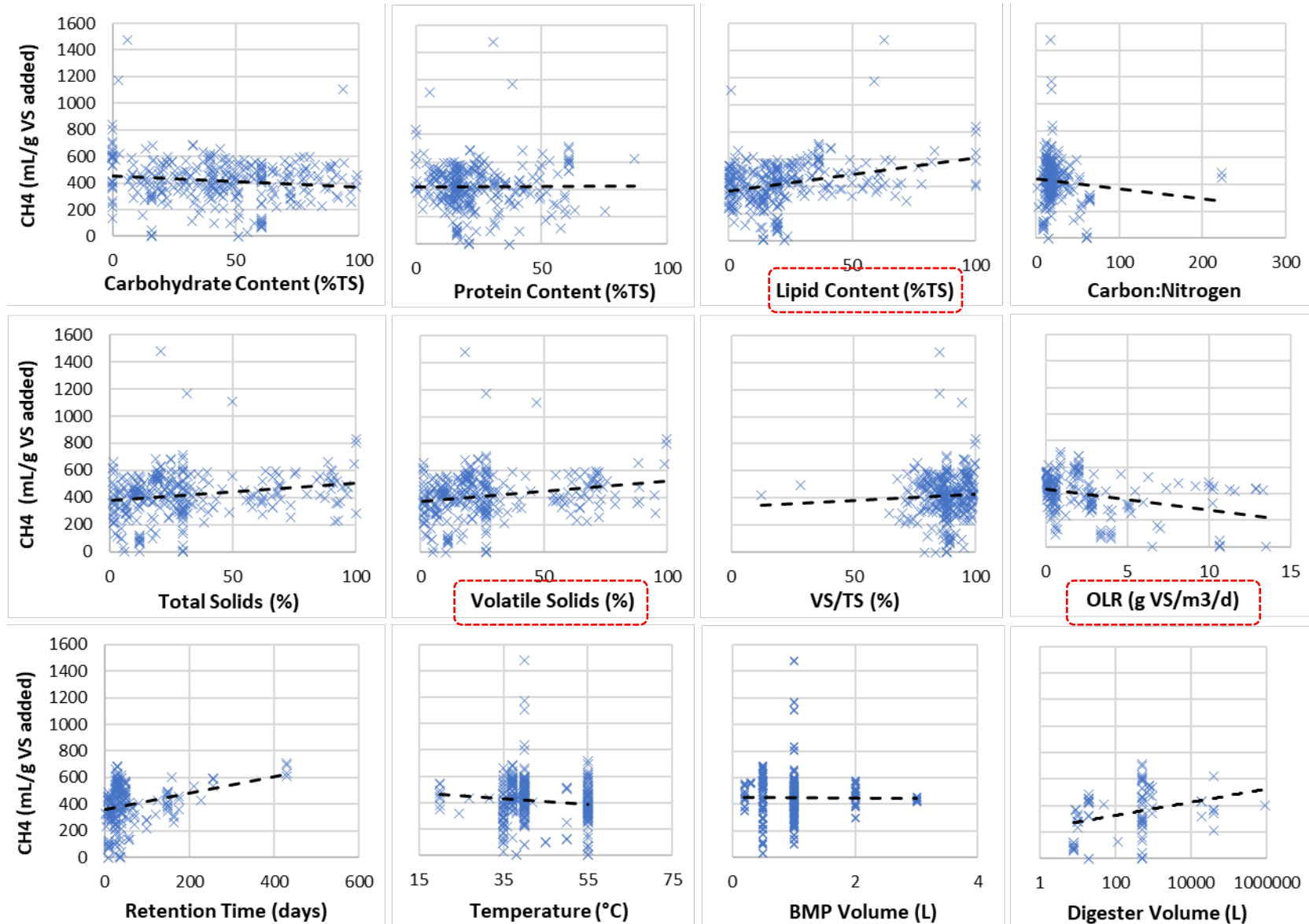
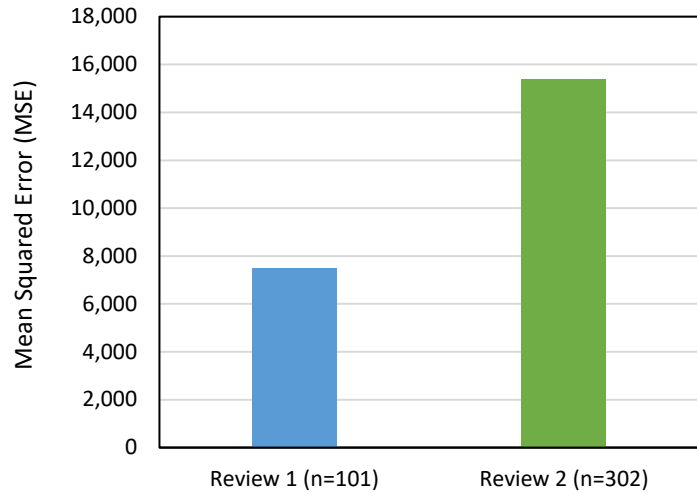
More data needed to improve predictability



Expanded dataset (n=302) resulted in even higher mean squared error

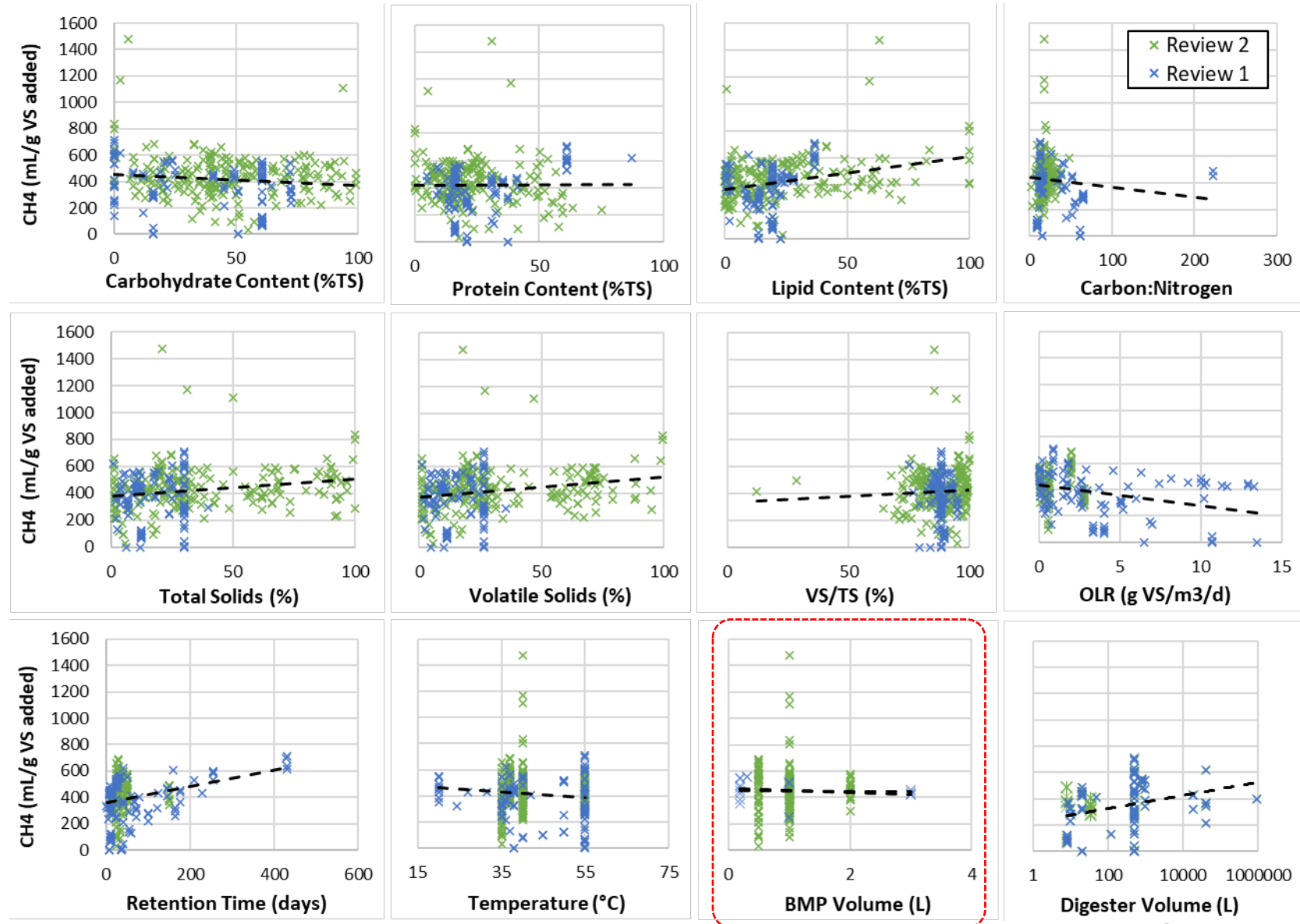
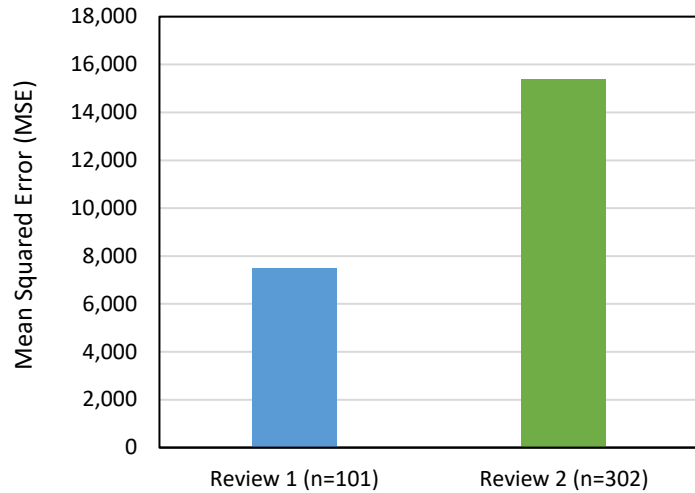
Key predictor variables included:

- Lipid Content
- Organic Loading Rate
- Volatile Solids



Expanded dataset included more small scale (≤ 3 L) biomethane potential data

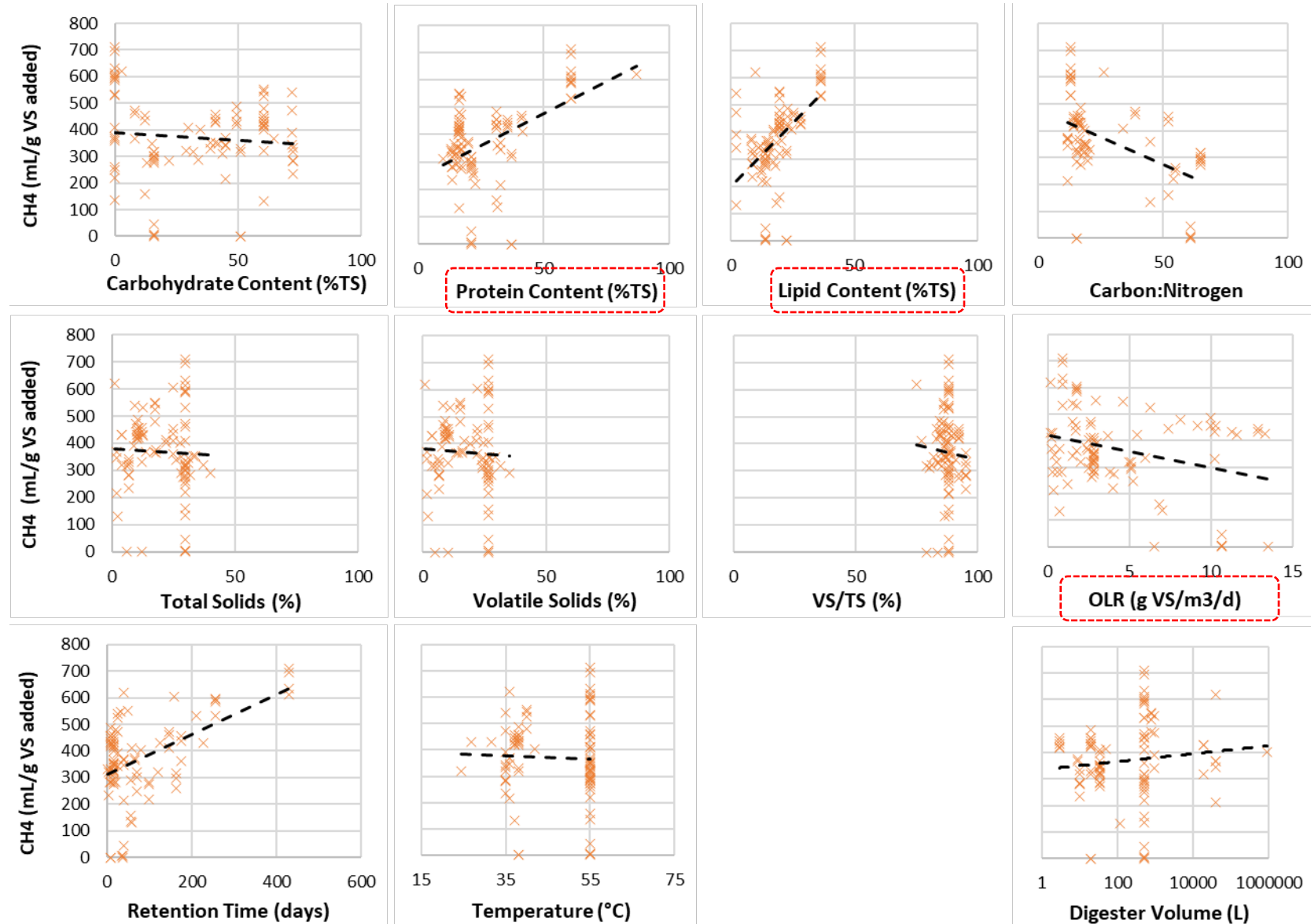
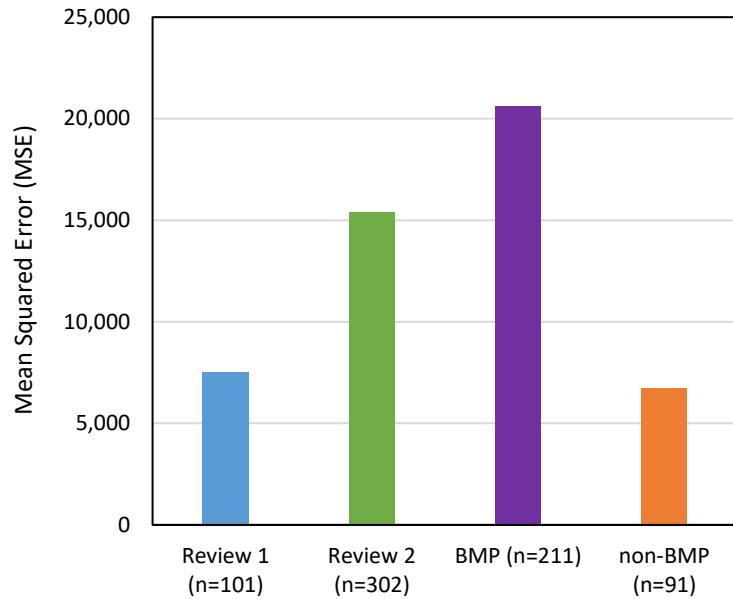
- **BMP literature had more description of feedstock characteristics**
- **Small scale BMP testing may not represent realistic and/or predictable outcomes**



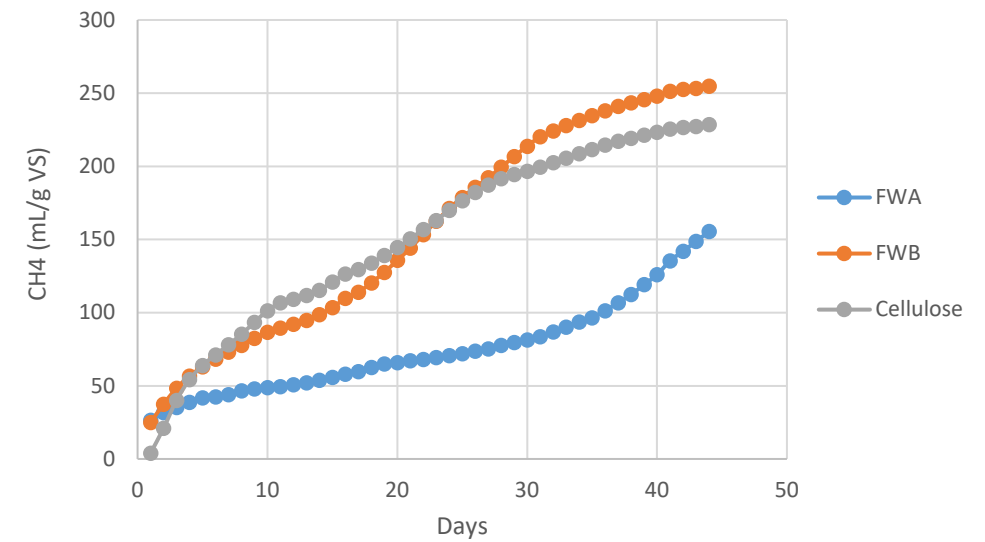
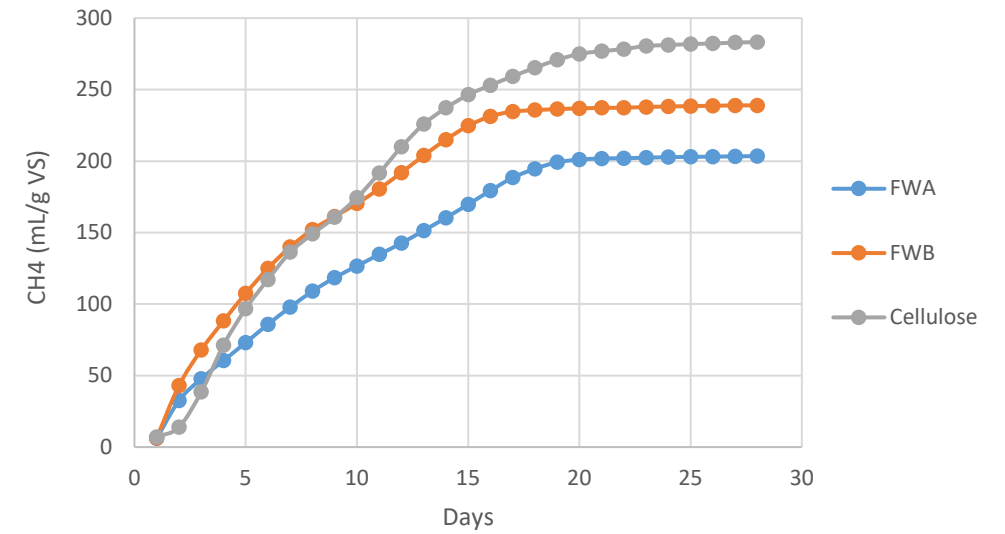
Removing BMP data resulted in lowest MSE – More data needed to improve

Key predictor variables included:

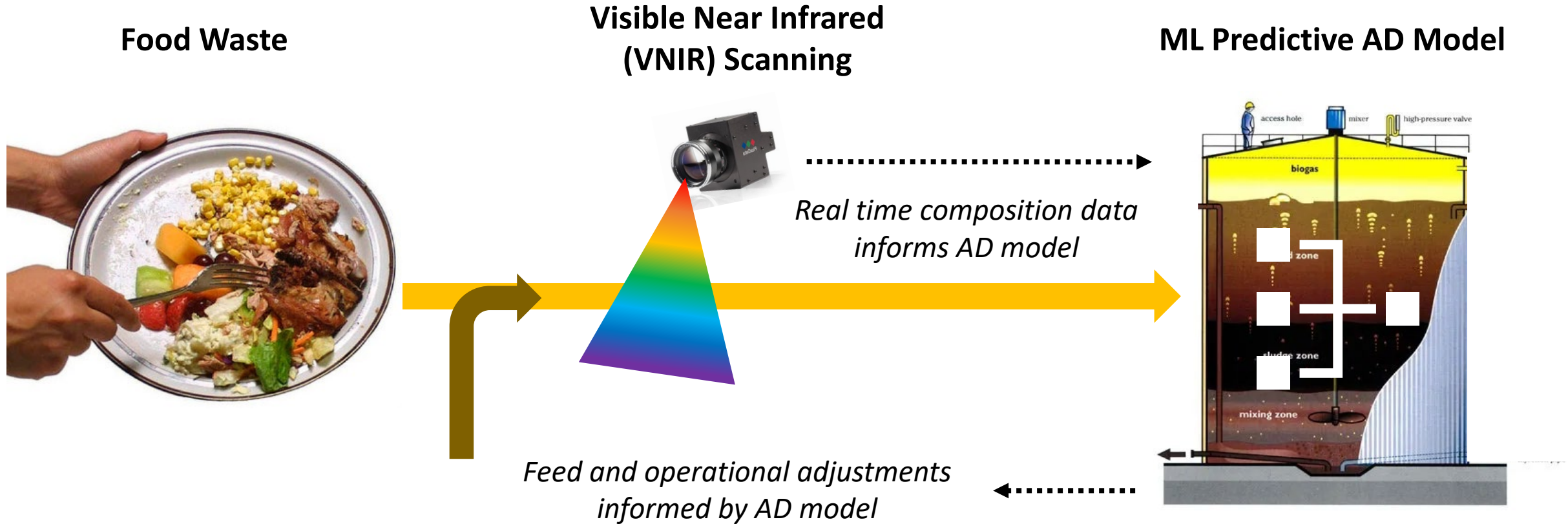
- Lipid Content
- Protein Content
- Organic Loading Rate



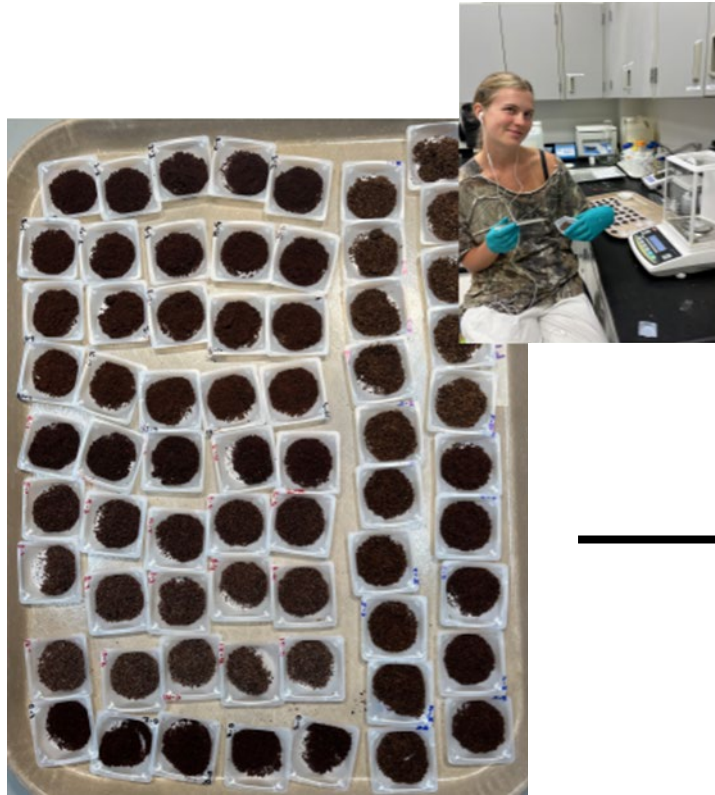
On-going Hinkley Project – Data collection from actual food waste



Long term goal: To develop a dynamic process control system



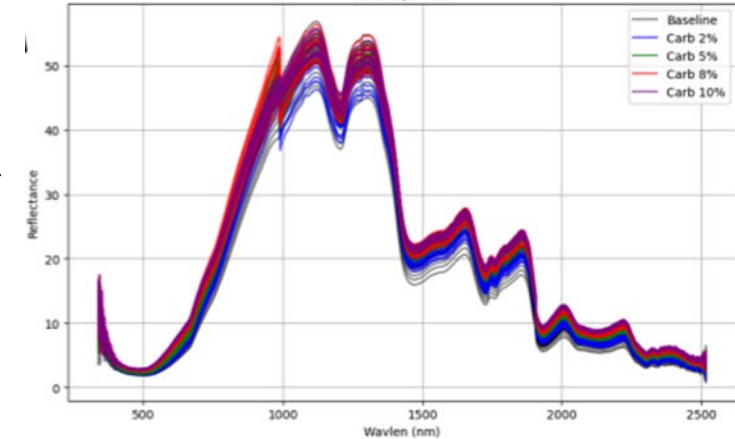
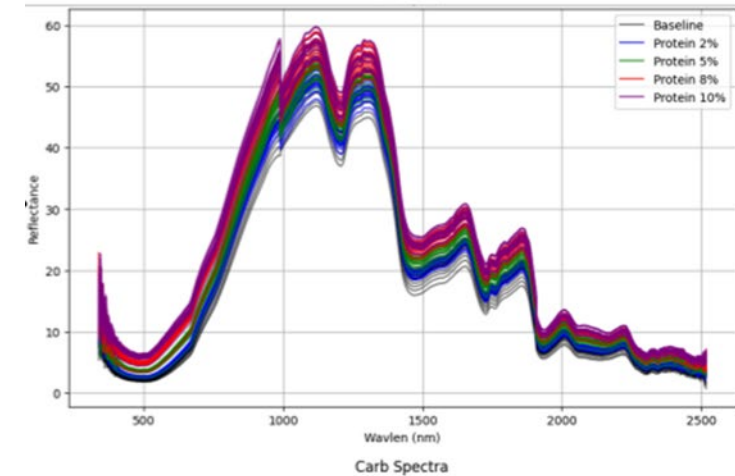
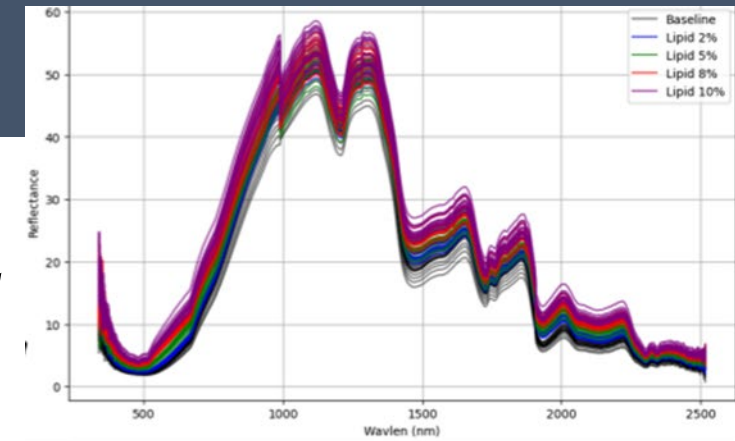
Using VNIR to characterize food waste composition



Collected VNIR spectral data (340-2500 nm) of spiked food waste samples

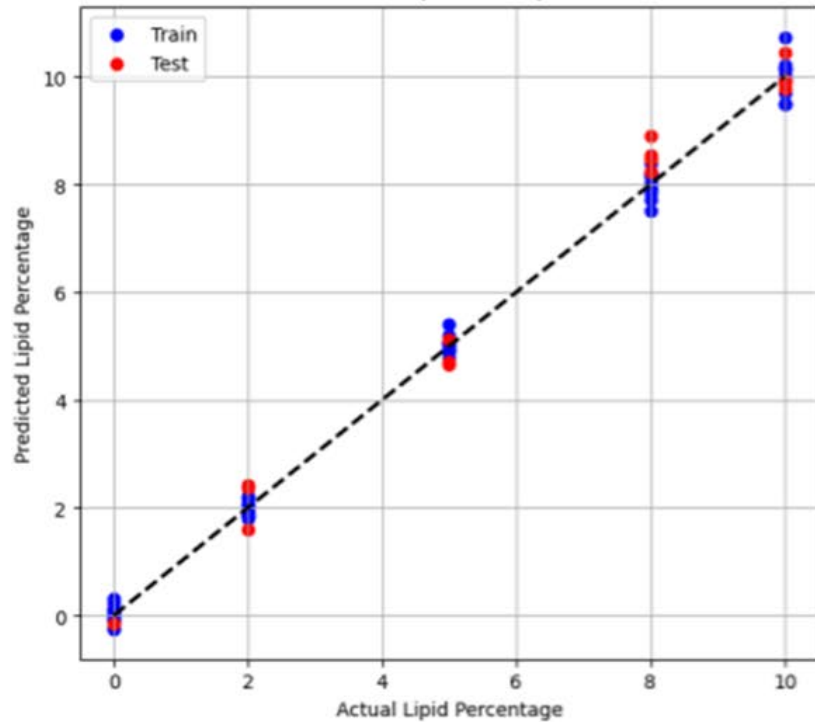
Spiked dried food waste samples with model compounds:

- corn starch (carb)
- collagen (protein)
- Triglyceride (lipid)



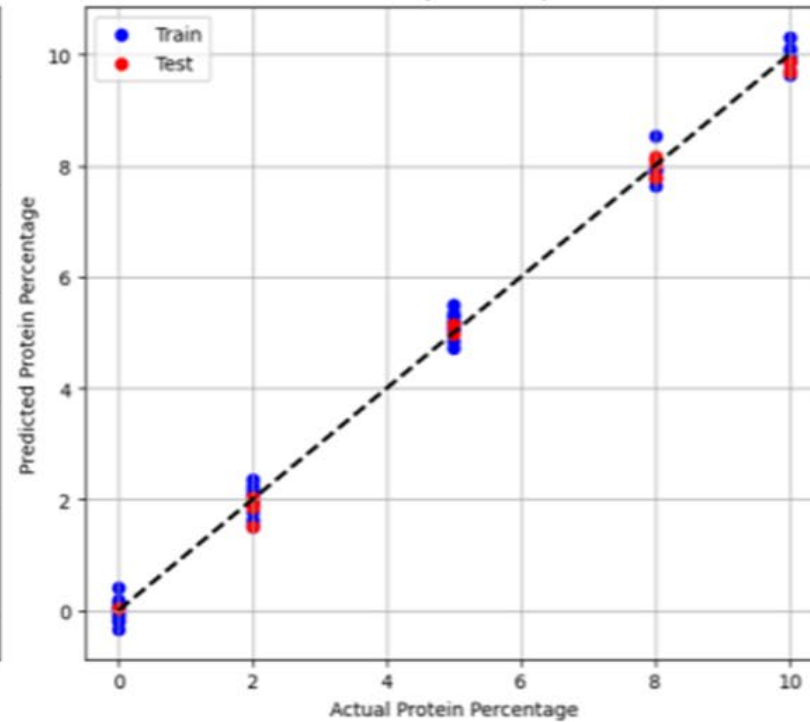
Lipid

Transform Snv Detrend Data - Lipid (5 components, RS: 28733)
Train: RMSE=0.22, R2=1.00, AIC=-189.27
Test: RMSE=0.39, R2=0.99, AIC=-38.04



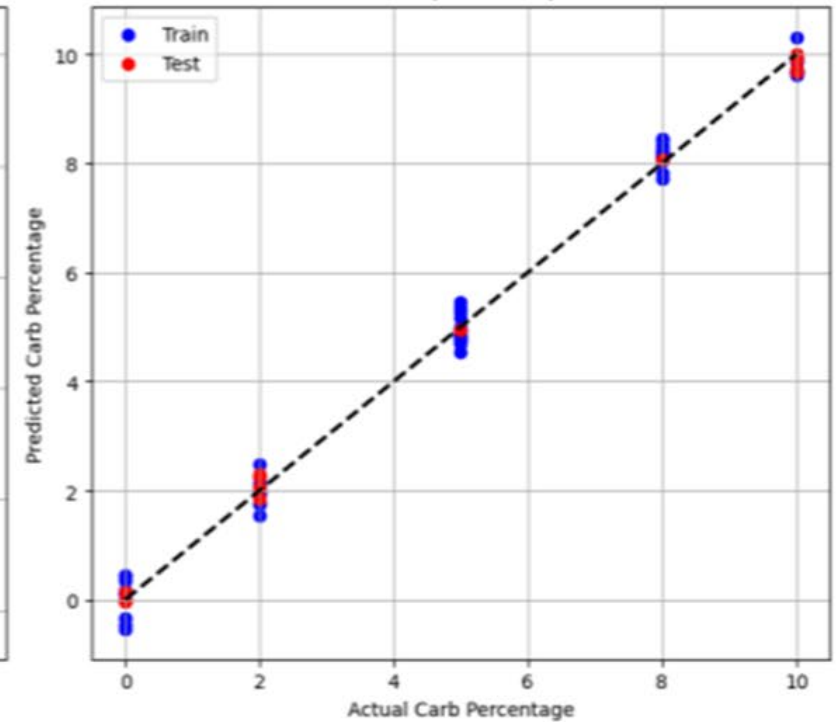
Protein

Transform Snv Detrend Data - Protein (4 components, RS: 17663)
Train: RMSE=0.22, R2=1.00, AIC=-188.31
Test: RMSE=0.19, R2=1.00, AIC=-57.56



Carbs

Transform Snv Detrend Data - Carb (5 components, RS: 50678)
Train: RMSE=0.27, R2=0.99, AIC=-165.83
Test: RMSE=0.17, R2=1.00, AIC=-63.88

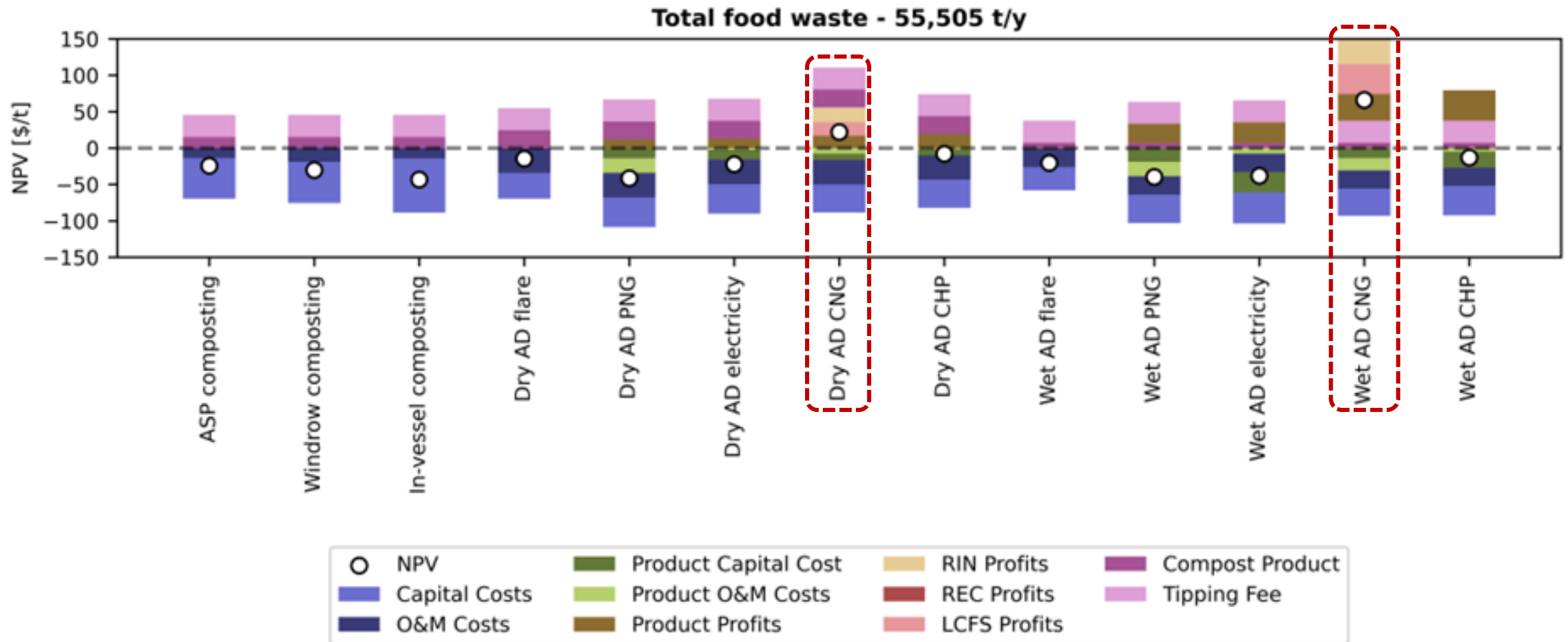


Macromolecule	n	RMSE	R ²
Lipids	5	0.39	0.99
Protein	4	0.19	1.00
Carbs	5	0.17	1.00

Ryan Lam
MS Graduate



Challenge 2: Economic Viability



Initial results from NREL analysis for the City of Gainesville estimating net present value (NPV) for community food waste composting and anaerobic digestion (AD) scenarios with various biogas upgrading options: PNG = Pipeline Renewable Natural Gas, CNG = Compressed Natural Gas, CHP = Combined Heat and Power.

City of Gainesville AD Feasibility Study



Co-digestion:

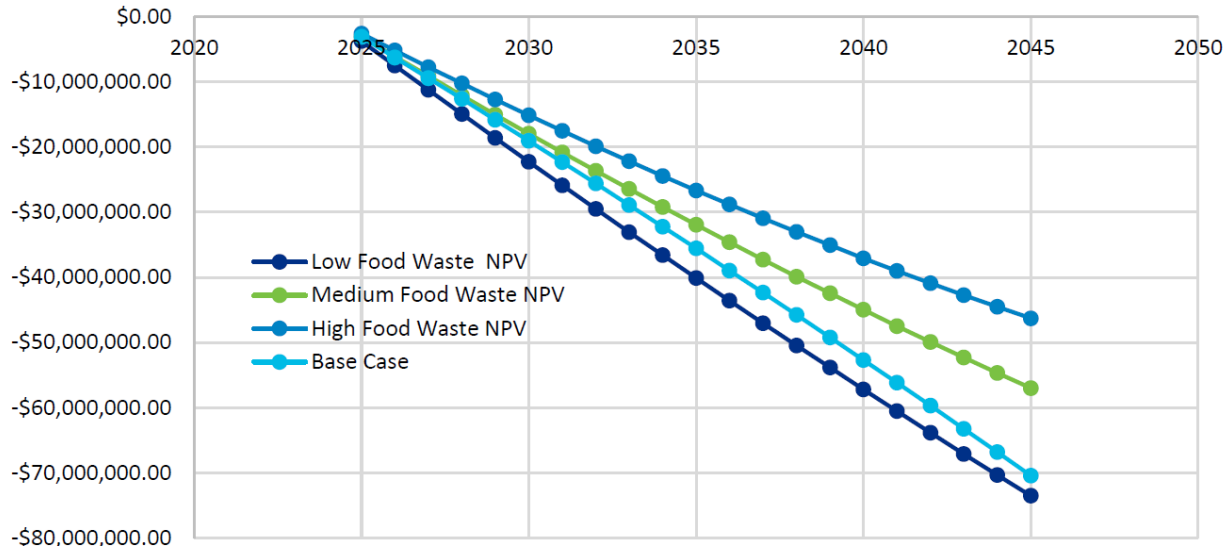


Zero Waste Ordinance:

EFFECTIVE: 6/1/2023

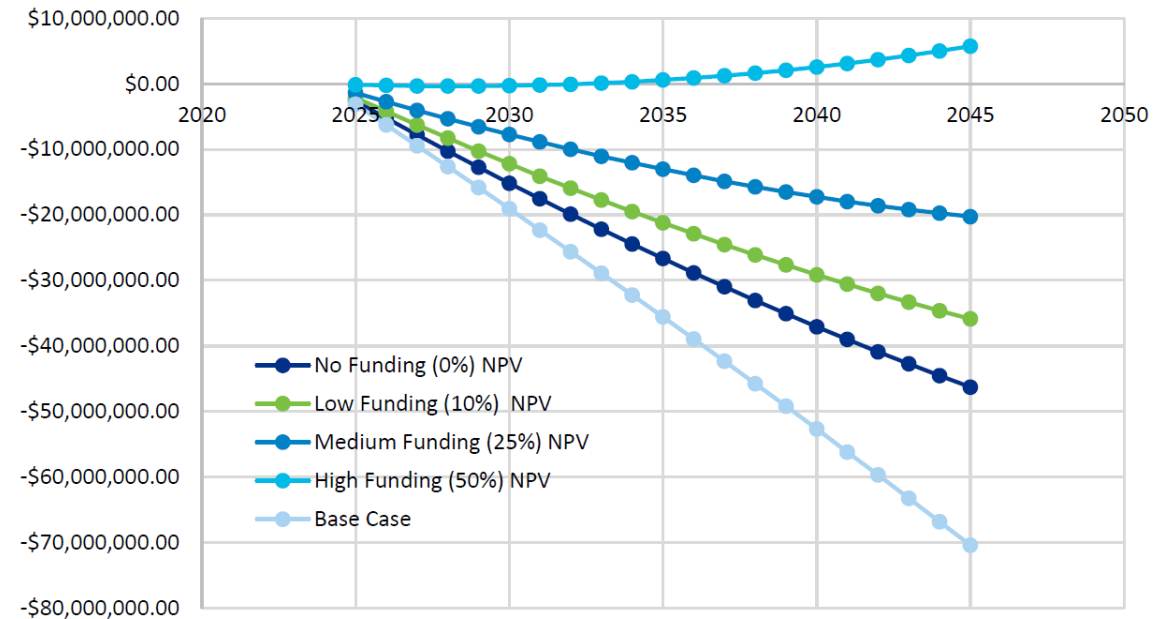
Commercial establishments that generate one cubic yard of food waste or more per week shall separate food waste from the waste stream and collect food waste in containers that are separate from garbage and recovered materials.

Effects of Food Waste Variability on Net Present Values



Low feedstock availability will reduce economic viability

Effects of Funding on Net Present Value



Grants or other funding needed to generate value

Challenge 3: Digestate Management



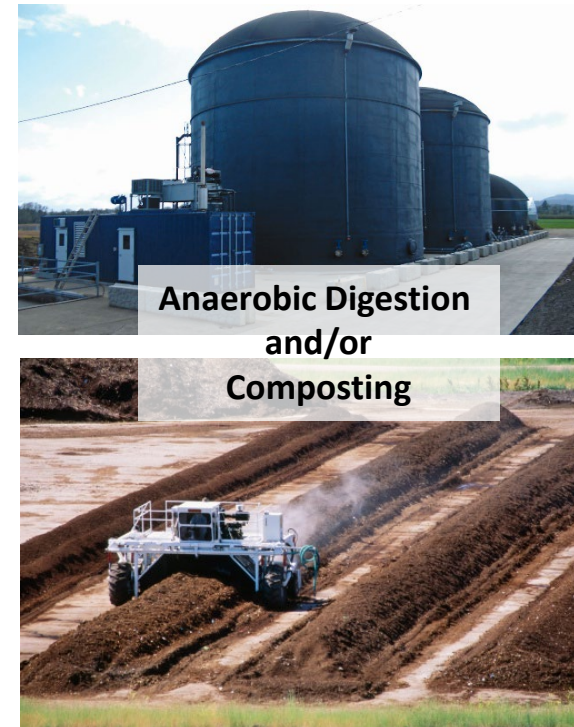
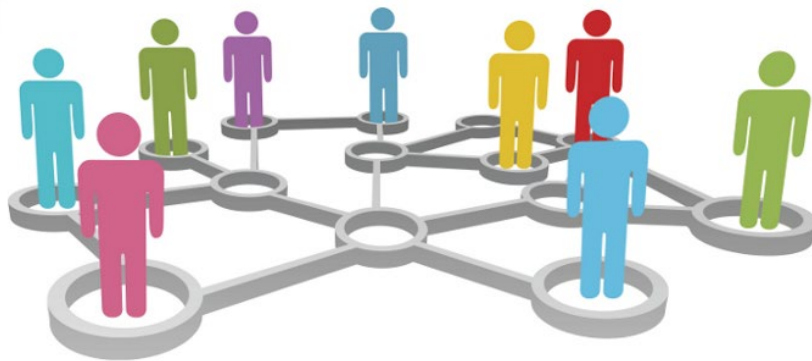
- Must make beneficial use of digestate (water and nutrients)
- Digestate dewatering/drying may be necessary
- Solids can be land applied, composted or upgraded to higher value product
- Partnerships/agreements need to be established

Current Hinkley Project:

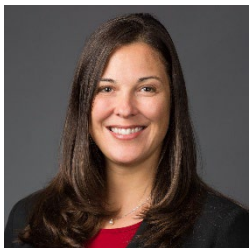
Evaluating Barriers to Adoption of Sustainable Food Waste Management

Interviewing Stakeholders:

- Policy Makers and Solids Waste Managers
- Restaurant Owners
- Composters



Identify barriers and opportunities including education/outreach



Dr. Catherine Campbell
Assistant Professor
Family Youth and Community Sciences

Thank you!

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UF Space Research Initiative

CDM Smith



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ENGINEERING

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