



31 MARCH 2022



15:00 CEST

IWA Specialist Group on Modelling and Integrated Assessment Webinar Series

State-of-the-art modelling on anaerobic mixed-culture fermentation

Speakers



Ulf Jeppsson
Lund University



Robbert Kleerebezem
Delft University of
Technology



Alberte Regueira
Universidade de
Santiago de Compostela



Adam Kovalovszki
DTU Environment



Matthew Scarborough
University of Vermont



The webinar is going to be recorded and shared on the MIA SG Youtube channel afterward.



MIA Welcome Note



IWA Modelling and Integrated Assessment Specialist Group

Dr. Ulf Jeppsson (Chair of MIA SG)

Dr. Elena Torfs (Vice-chair of MIA SG)



inspiring change



MODELLING AND INTEGRATED ASSESSMENT SPECIALIST GROUP (MIA SG)



*“This group targets people from research, consulting companies, institutions and operators to think along **the use of models and computing tools to support the understanding, management and optimization of water systems.**”*

PRIORITIES

- Interact with other IWA SGs and other professional organizations
- Organize specialized conferences, sessions and workshops
- Engage and activate YWPs in the domain.

CURRENTLY 1900 MEMBERS

How to find us



Website: <http://iwa-mia.org/>



Website: <http://iwa-mia.org/>

<https://iwa-connect.org>



MIA SG: ACTIVITIES

Task Groups (TGs)

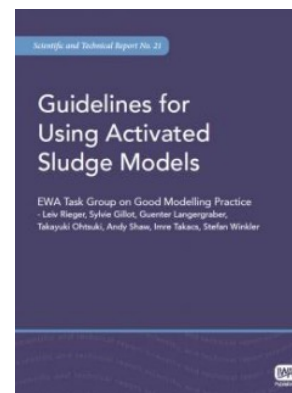
- Benchmarking of Control Strategies for WWTPs (BSM) **AND** Good Modelling Practice (GMP) **AND** Design and Operations Uncertainty (DOUT) (**All three finished**)
- Generalised Physicochemical Modelling (PCM) (**almost done**)
- Use of Modelling for Minimizing GHG Emissions from Wastewater Systems (GHG) (**almost done**)
- Membrane Bioreactor Modelling and Control (MBR)
- Good Modelling Practice in Water Resource Recovery Systems

Working Groups (WGs)

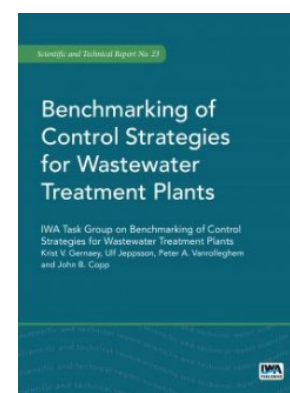
- Integrated Urban Water Systems (IUWS)
- Computational Fluid Dynamics (CFD)
- Good Modelling Practice (GMP)

Conferences / Events

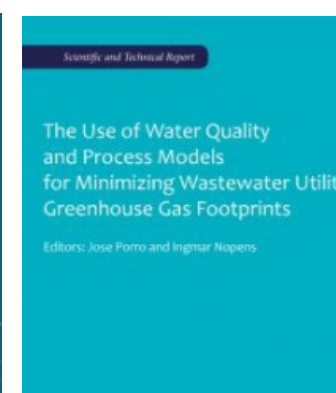
- WRRmod
- Watermatex



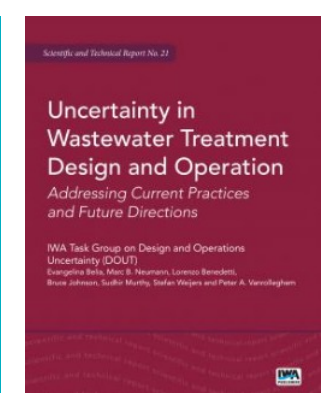
STR
(Sept. 2012)



STR
(Sept. 2014)



STR
(2022)



STR
(2022)



MIA SG: UPCOMING CONFERENCES

8th Water Resource Recovery Modelling seminar (WRRmod2022+)

- Location: Stellenbosch, South Africa, 15-18 January 2023
- Chair: Dr. David Ikumi (Univ. Cape Town)



11th Symposium on Modelling and Integrated Assessment (Watermatex2023)

- Location: Québec City, Canada, late summer 2023
- Chair/vice-chair: Prof. Peter Vanrolleghem (Univ. Laval)/Dr. Elena Torfs (Univ. Ghent)



9th Water Resource Recovery Modelling seminar (WRRmod2024), Stowe, Vermont, USA

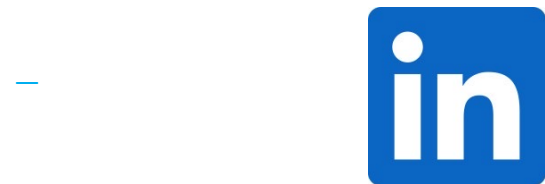


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Follow the Modelling and Integrated Assessment Specialist Group on:



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<https://www.linkedin.com/company/iwa-mia-specialist-group-on-modelling-and-integrated-assessment>



https://twitter.com/iwa_mia_sg




<http://iwa-mia.org>

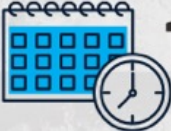
to get informed about our latest events, publications and news!

MIA Open Group meeting at WWC&E2022, Copenhagen, Denmark

INTRODUCTION TO THE WEBINAR






31 MARCH 2022
 **15:00 CEST**


IWA Specialist Group on Modelling and Integrated Assessment Webinar Series

State-of-the-art modelling on anaerobic mixed-culture fermentation


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
Alberte Regueira
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Matthew Scarborough
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Modeling Mixed Culture Fermentations

A short introduction

Robbert Kleerebezem (R.Kleerebezem@tudelft.nl)

Delft University of Technology



Modeling *Wild* Fermentations

A short introduction

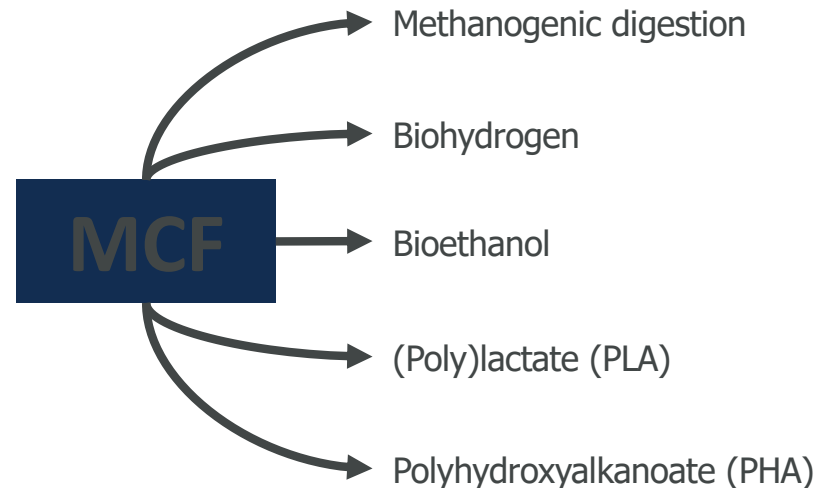
WHY MIXED CULTURE FERMENTATION (MCF)?



Produce building blocks for other processes

Substrate: Carbohydrates, Glycerol...

Products: Acetate, Propionate, Butyrate, Ethanol, Lactate, Hydrogen...

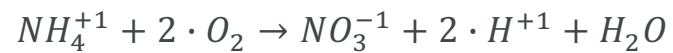


WHY MCF IS INTERESTING?



Thermodynamics drive microbial conversions:

The environment selects for maximization of energy harvesting



WHY MCF IS INTERESTING?

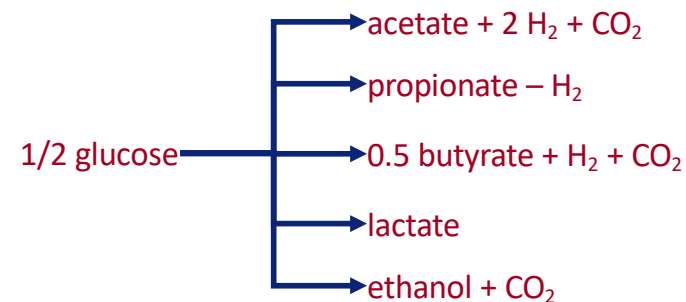


Thermodynamics drive microbial conversions:

The environment selects for maximization of energy harvesting



But what determines the end-product of carbohydrate fermentations?



ANAEROBIC CARBOHYDRATE FERMENTATION



Cabbage



Sauerkraut



Grapes



Wine



Potato peel



Butyric acid



HISTORY OF MODELLING MCF



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Printed in Great Britain.

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MATHEMATICAL MODELLING OF THE ANAEROBIC DIGESTION PROCESS: REGULATORY MECHANISMS FOR THE FORMATION OF SHORT-CHAIN VOLATILE ACIDS FROM GLUCOSE

F. E. Mosey

Water Research Centre, Elder Way, Stevenage, Herts SG1 1TH, U.K.

ABSTRACT

A model of the anaerobic digestion process is presented which attempts to explain the complex patterns of volatile acid production in the anaerobic digestion process. The hydrogen-utilising methane bacteria are identified as

HISTORY OF MODELLING MCF

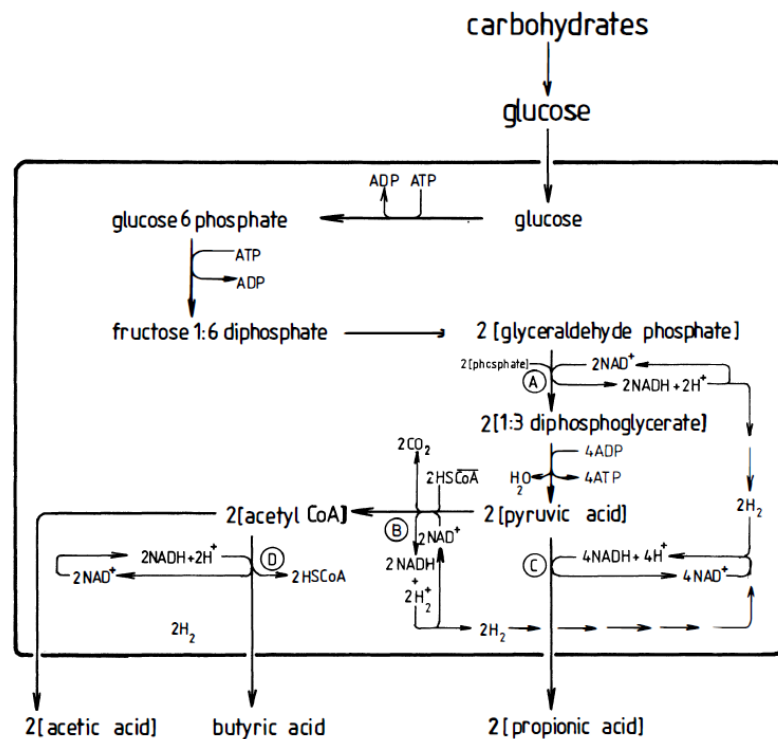


Fig. 2. Metabolic pathways inside acid-forming bacteria

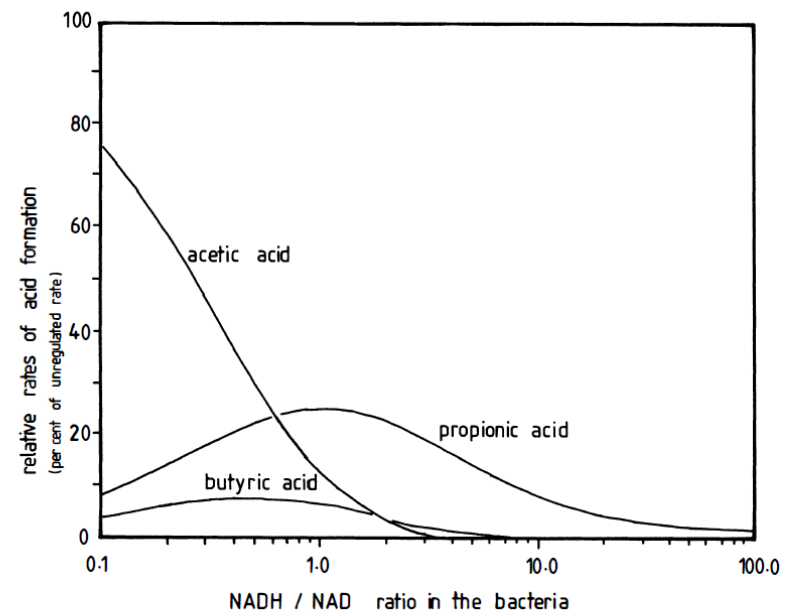


Fig. 3. Relative rates of acid-formation at different NAD^+/NADH ratios as predicted by these rate equations



LECTURES OF TODAY!

- **Alberte Regueira**

Bioenergetic modelling for predicting the stoichiometry of AF

- **Adam Kovalovszki**

*Inferring metabolic interactions from genome-scale data: anaerobic digestion
'under the microscope'*

- **Matthew Scarborough**

Guild-based metabolic modeling for deciphering anaerobic bioprocesses



AGENDA AND HOUSEKEEPING

Speaker 1

Alberte Regueira (*Universidade de Santiago de Compostela, Spain*)

Speaker 2

Adam Kovalovszki (*Technical University of Denmark*)

Speaker 3

Matthew Scarborough (*University of Vermont, USA*)

Q&A Session Moderator: Robbert Kleerebezem (*Delft University of Technology, The Netherlands*)

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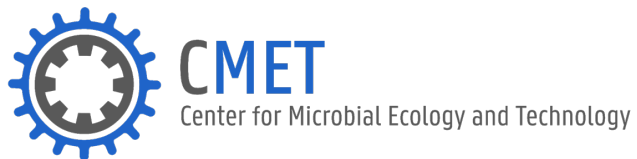
BIOENERGETIC MODELLING FOR PREDICTING THE STOICHIOMETRY OF ANAEROBIC FERMENTATION

Alberte Regueira (Alberte.RegueiraLopez@UGent.be)
Universidade de Santiago de Compostela



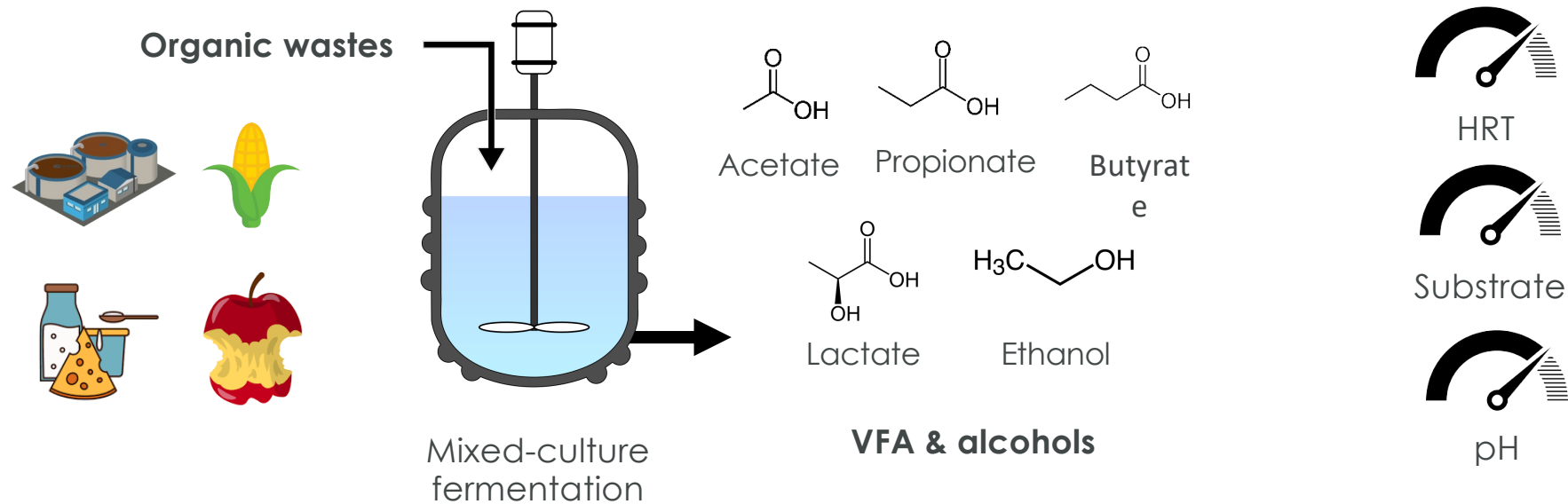
Bioenergetic modelling for predicting the stoichiometry of anaerobic fermentation

Alberte Regueira López



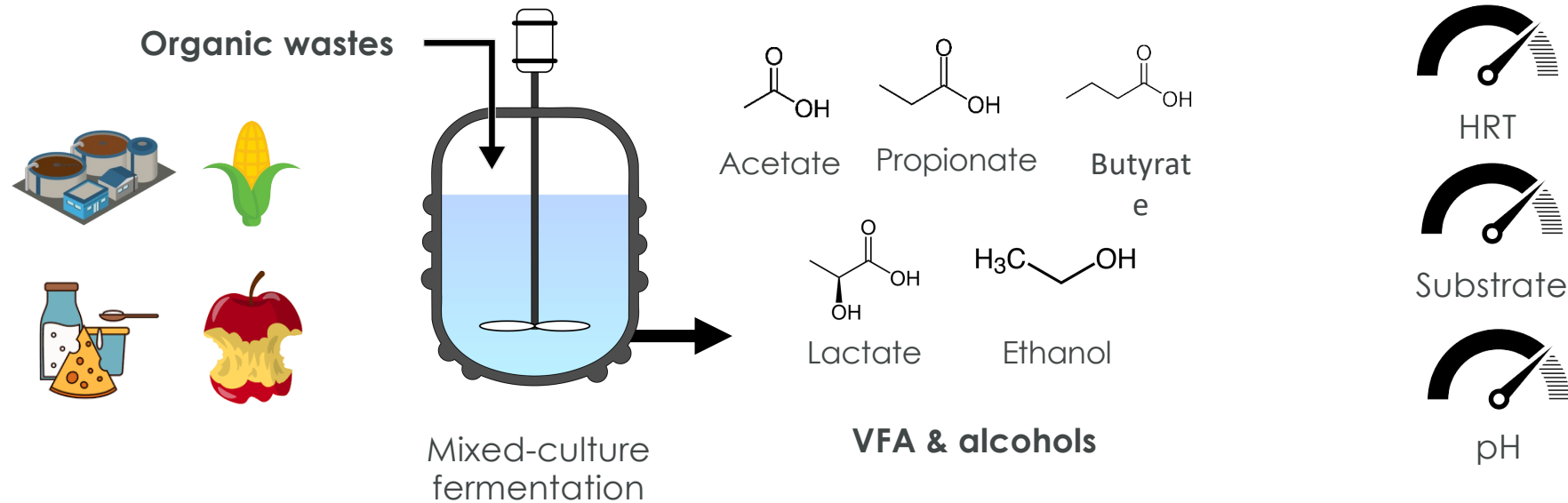


Anaerobic fermentation for VFA production





Anaerobic fermentation for VFA production



$$\Delta G = \Delta G^0 + RT \ln \frac{[P_1] \cdot [P_2]}{[S_1] \cdot [S_2]}$$

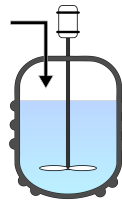
Standard value Environmental conditions



What are bioenergetic models?

Kinetic unstructured models

- ✓ Biomass is a black box
- ✓ Solve the macroscopic mass balances
- ✓ Variable selectivity is not addressed

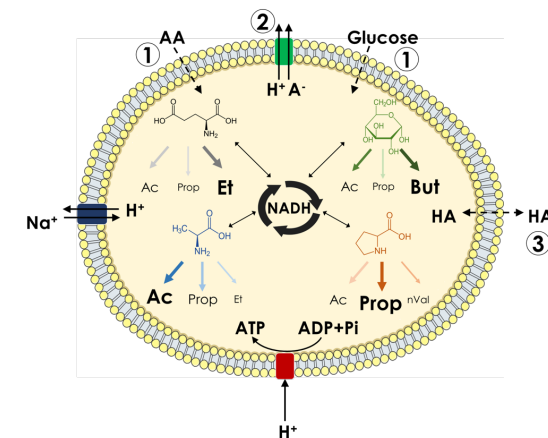


$$\frac{dC_i}{dt} = D \cdot (C_i^{IN} - C_i) + r_{j,i}$$

$$r_{j,i} = r_{max,j} \cdot \frac{C_i}{C_i + K_{j,i}} \cdot X$$

Bioenergetic models

- ✓ Intracellular processes are modelled
- ✓ Cell-environmental interactions
- ✓ Their task is limited to predict the process stoichiometry





The microbial community is modelled as an enzyme soup

Reality: **Multiple** species performing different or similar metabolic functions

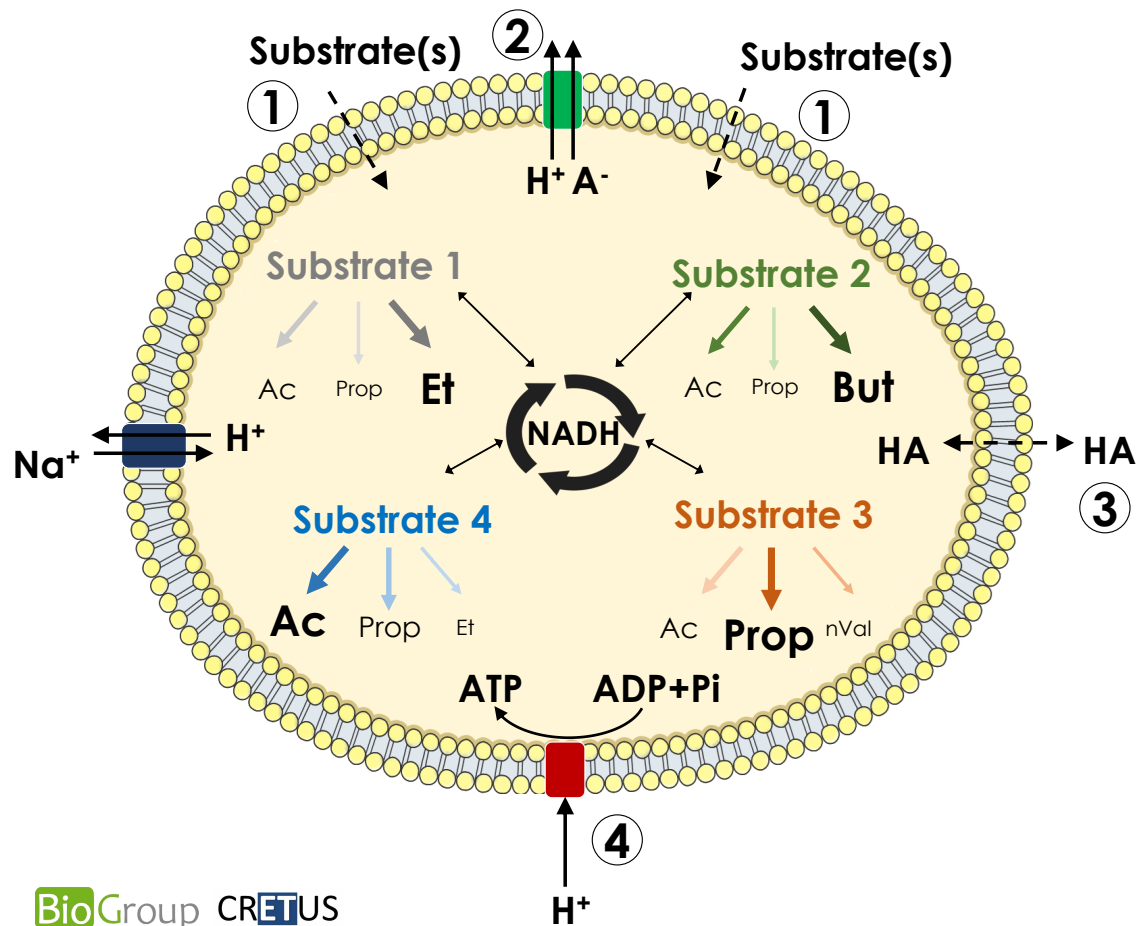


Model: **One** virtual species is able of performing **all** the metabolic functions of the community





One virtual microorganism does all the possible processes

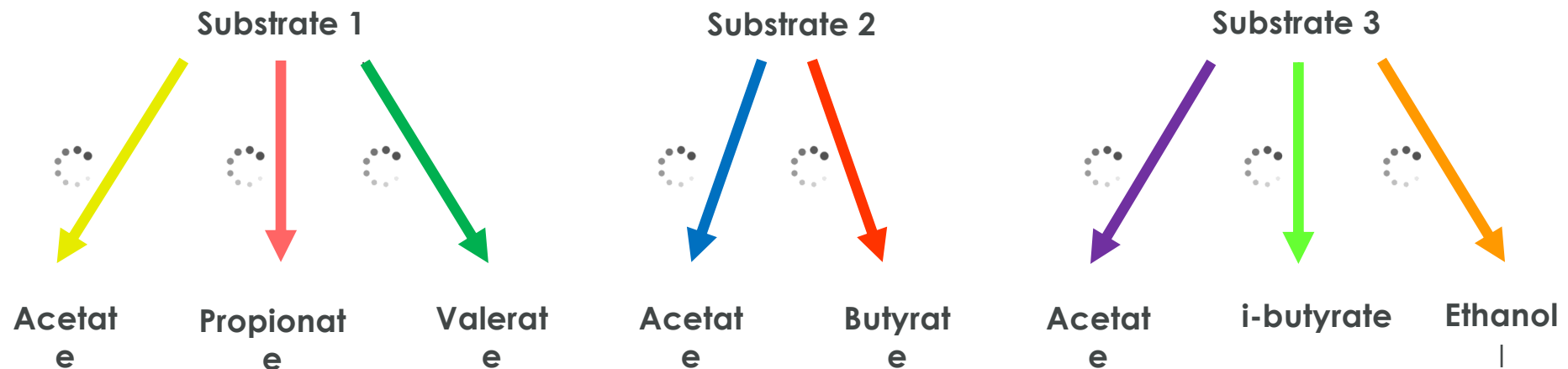


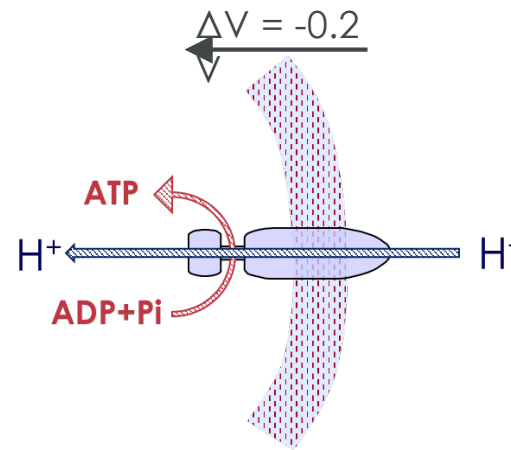
1. Substrate transport
2. Active transport of products
3. Passive transport of products
4. Energy conservation via proton motive force



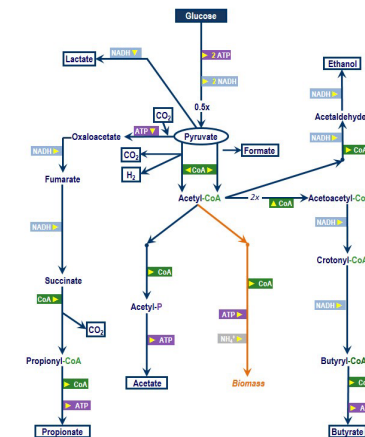
Flux balance analysis to determine product selectivity

- FBA determines the metabolite flow through the pathways of the metabolic network
- The flow distribution maximises a given objective (e.g. maximum growth rate)



$$r_{\text{ATP}} = r_{\text{TRANSPORT}} + r_{\text{PMF}} + r_{\text{CAT}}$$


Proton translocations



Catabolic ATP



Bioenergetic models are dynamic flux balance analysis models

$$\frac{dC}{dt} = D \cdot (C^{in} - C) + r(z, C)$$

$$z(t) / \max_z r_{ATP}(C(t))$$

Maximise ATP production



Intracellular conditions can also vary

s.t.

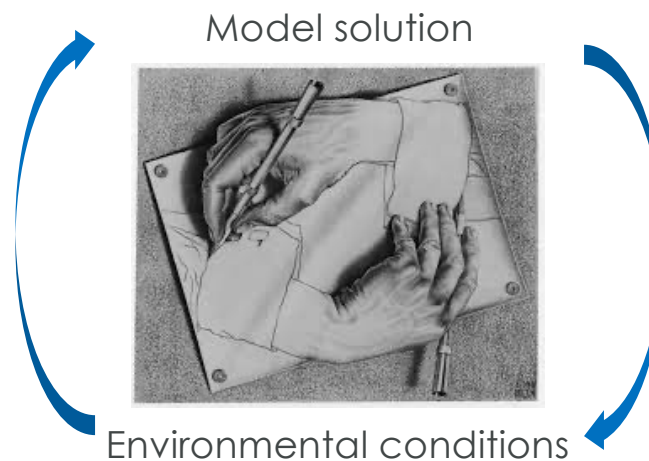
$$r_{NADH}(z) = 0$$

NADH is conserved

$$g(z) \leq 0$$

Environmental conditions

$$h(z) = 0$$





Bioenergetic models in literature

- Glucose
 - Rodríguez *et al.* (2006): First bioenergetic model
 - Zhang *et al.* (2013): Modifications on Rodríguez's model
 - González-Cabaleiro *et al.* (2015): State-of-the-art approach
- Protein
 - Regueira *et al.* (2020): First bioenergetic model for protein fermentation
- Protein and glucose cofermentation
 - Regueira *et al.* (2020): First bioenergetic model for co-fermentation

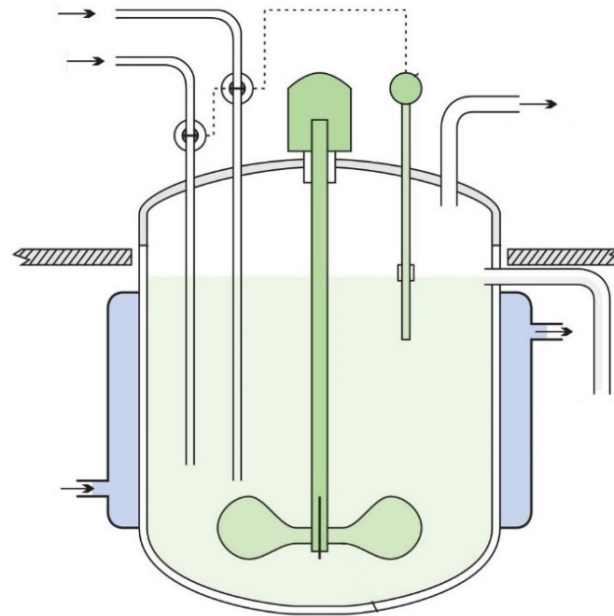
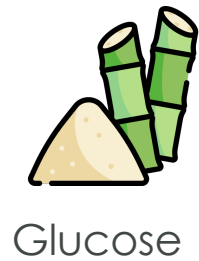


Bioenergetic models in literature

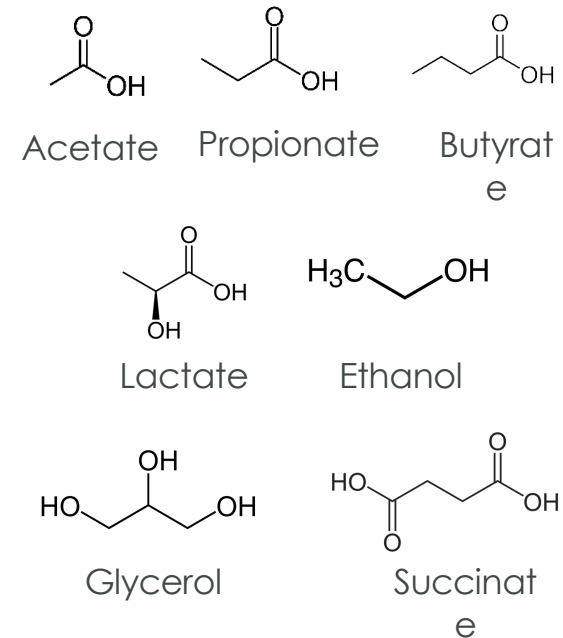
- Glucose
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Bioenergetic model for glucose fermentation

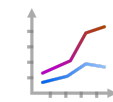
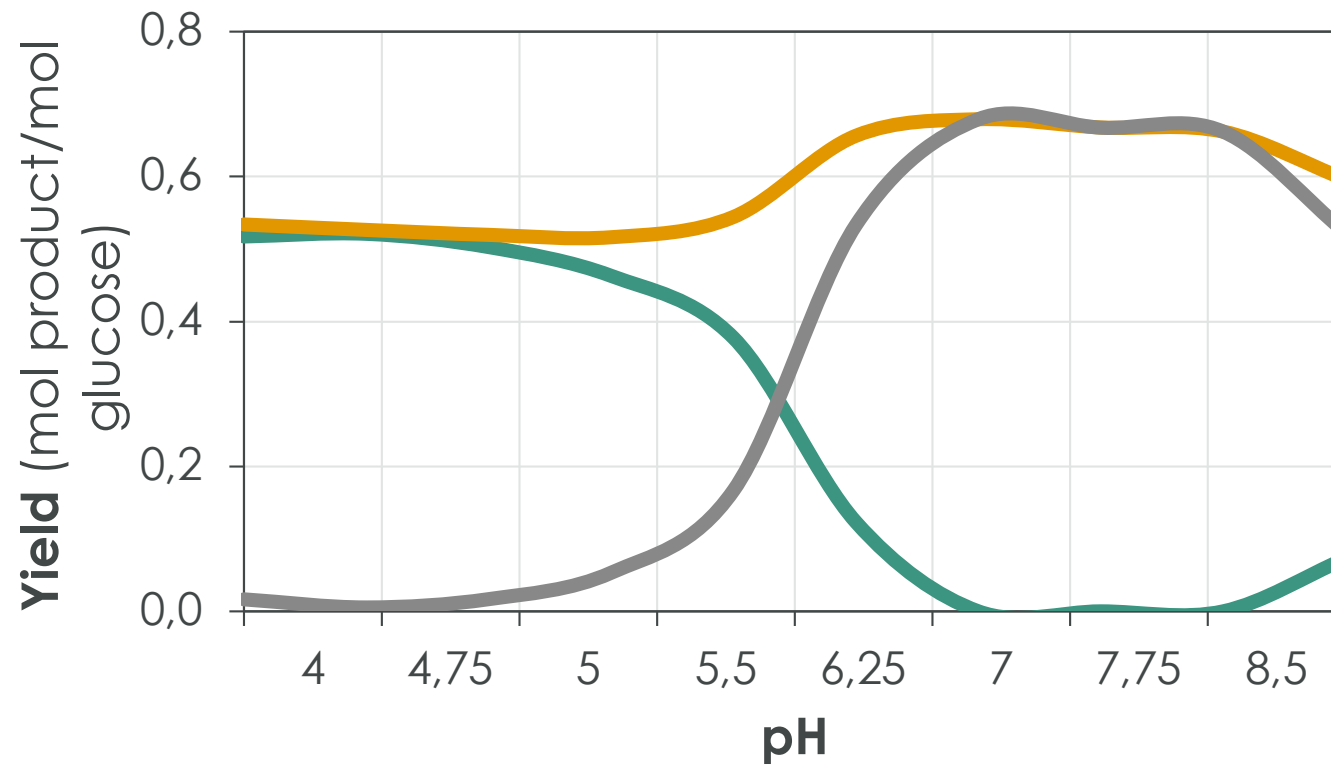


pH: 4-
8.5





The model predicts a shift with pH



Model results

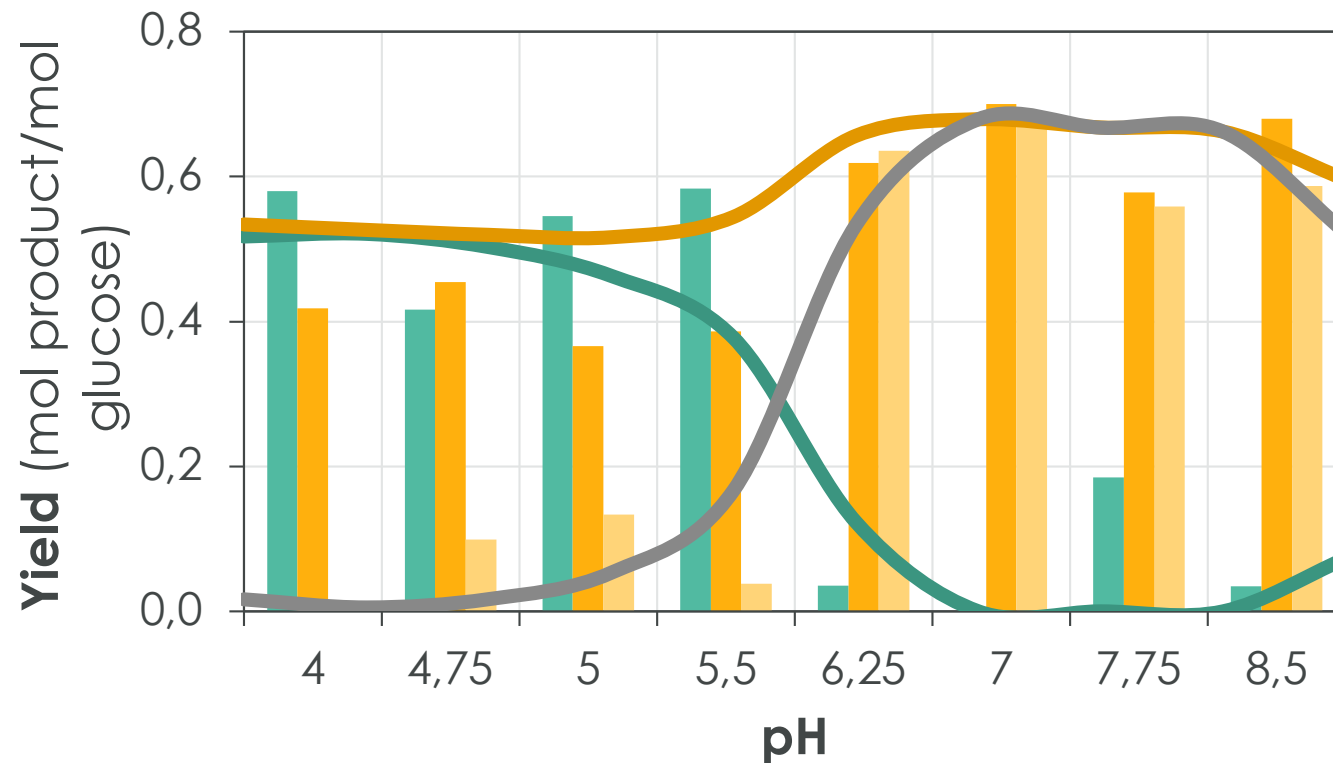
Lines

■ Butyrate ■ Acetate ■ Ethanol

Model results: González-Cabaleiro et al. (2015), PLoS ONE
Experimental results: Temudo et al. (2007), Biotechnol.
Bioeng.



The model was validated with experimental results

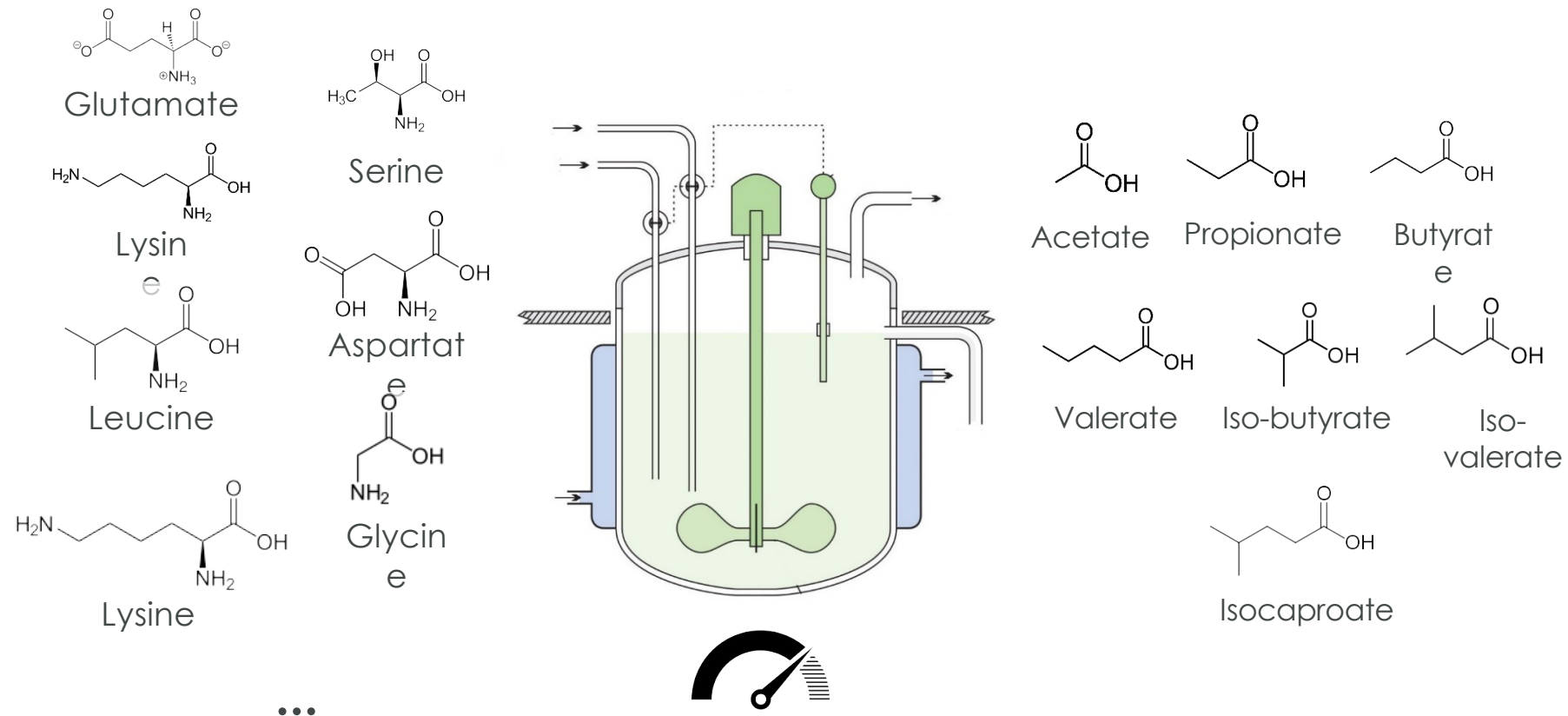


■ Butyrate ■ Acetate ■ Ethanol

Model results: González-Cabaleiro et al. (2015), PLoS ONE
 Experimental results: Temudo et al. (2007), Biotechnol.
 Bioeng.



Bioenergetic model for protein fermentation

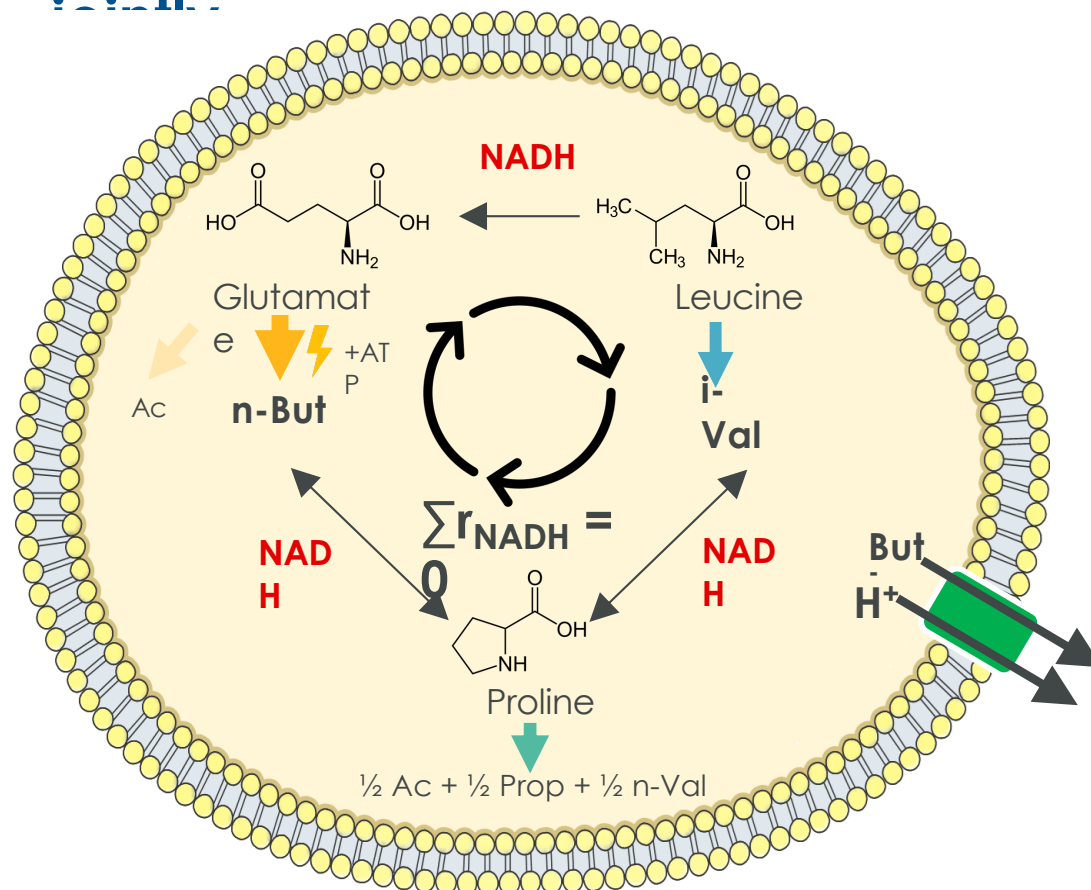


pH: 4-9



Enzyme soup approach: amino acids are converted

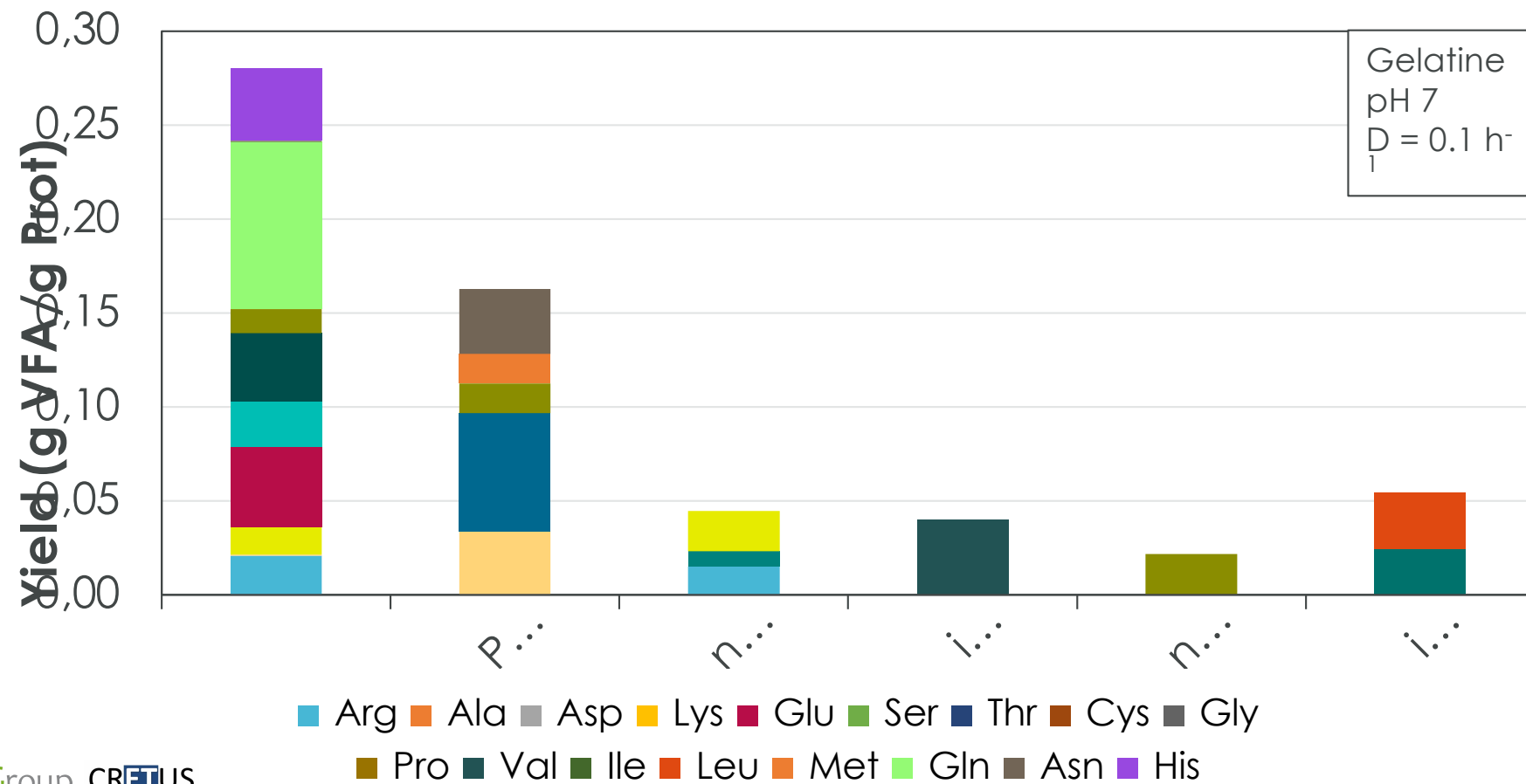
Enzymes



- Amino acids **interact** and **share** common metabolites (e.g. NADH) and affect common processes (e.g. VFA transport)



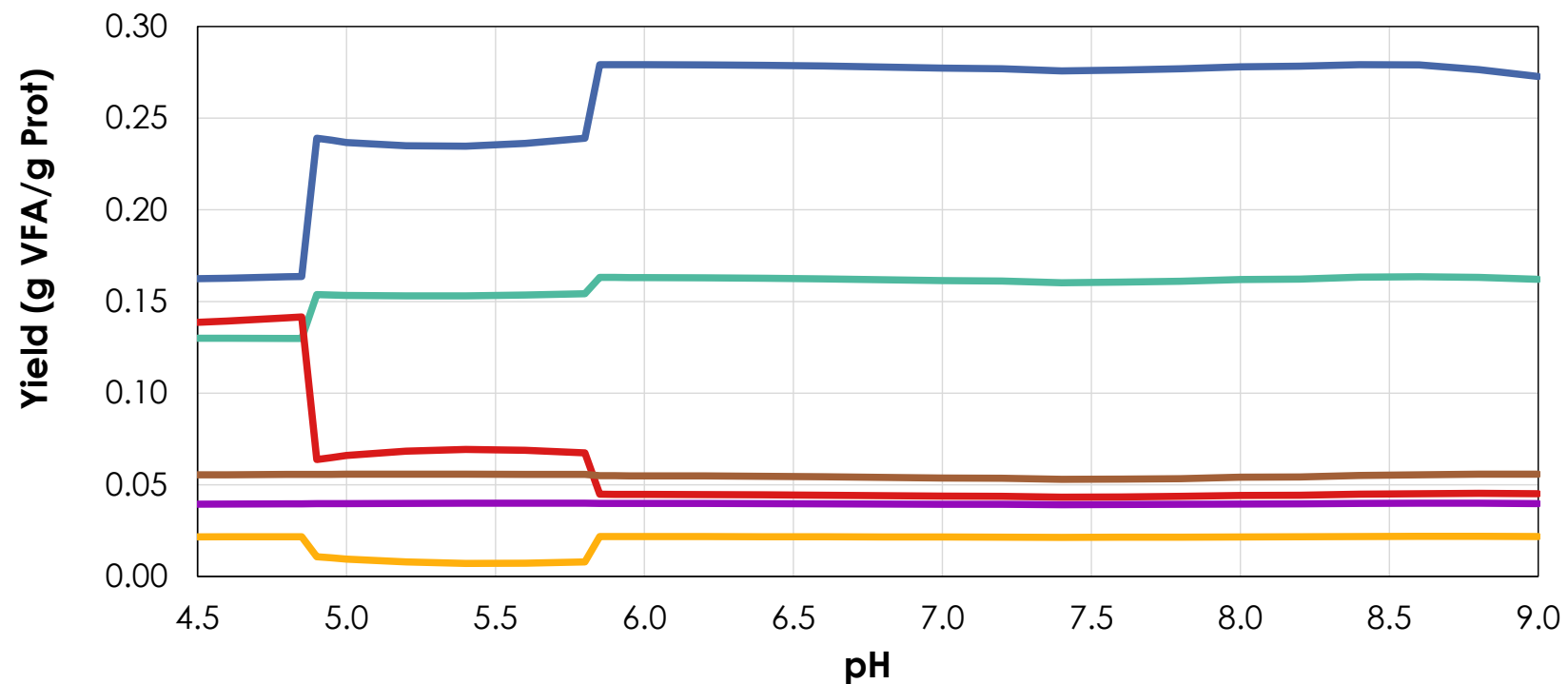
Predicted VFA yields from gelatine fermentation at pH 7





The product spectrum is sensitive to pH

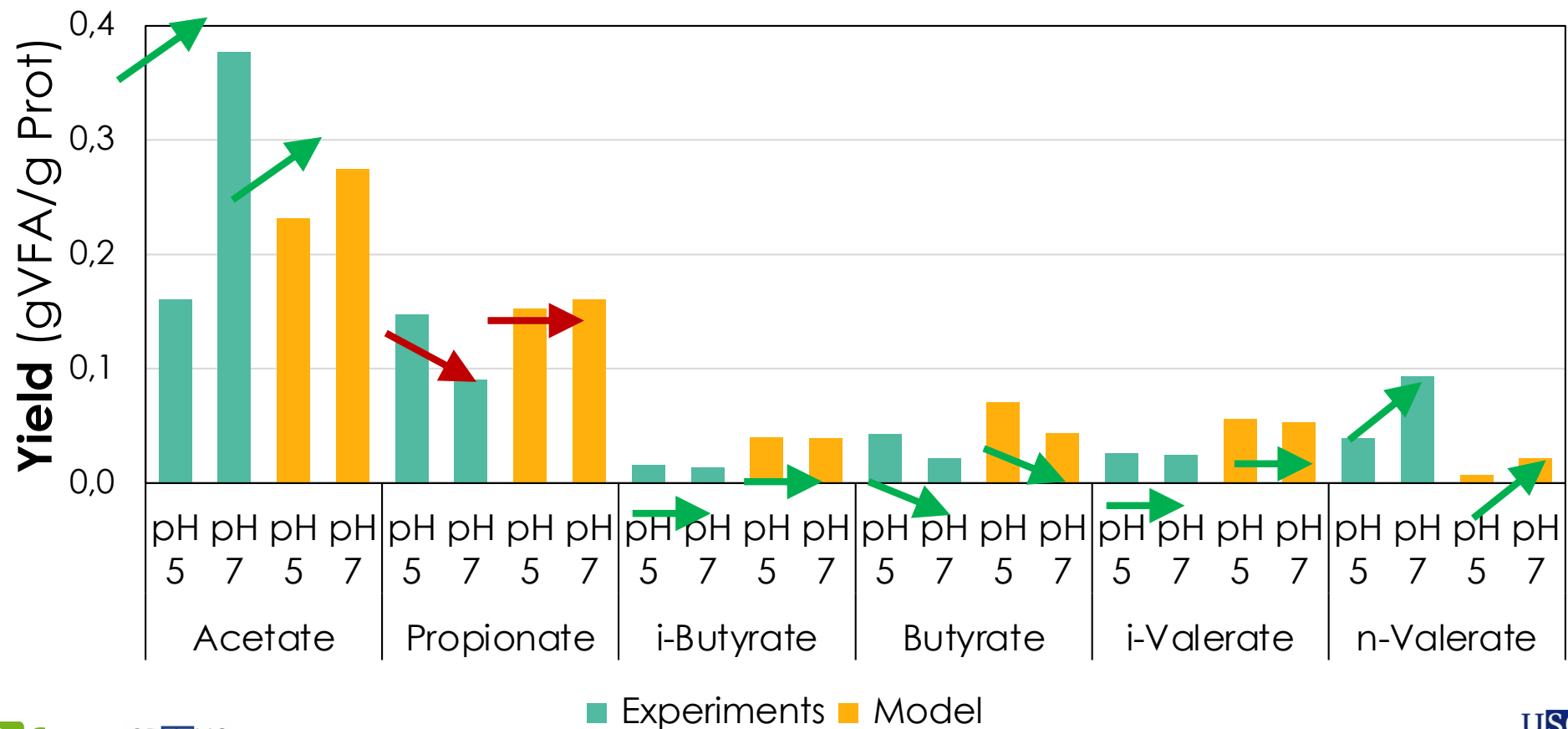
- Higher proton motive force energy available at low pH favours butyrate production



acetate propionate n-butyrate i-butyrate n-valerate i-valerate

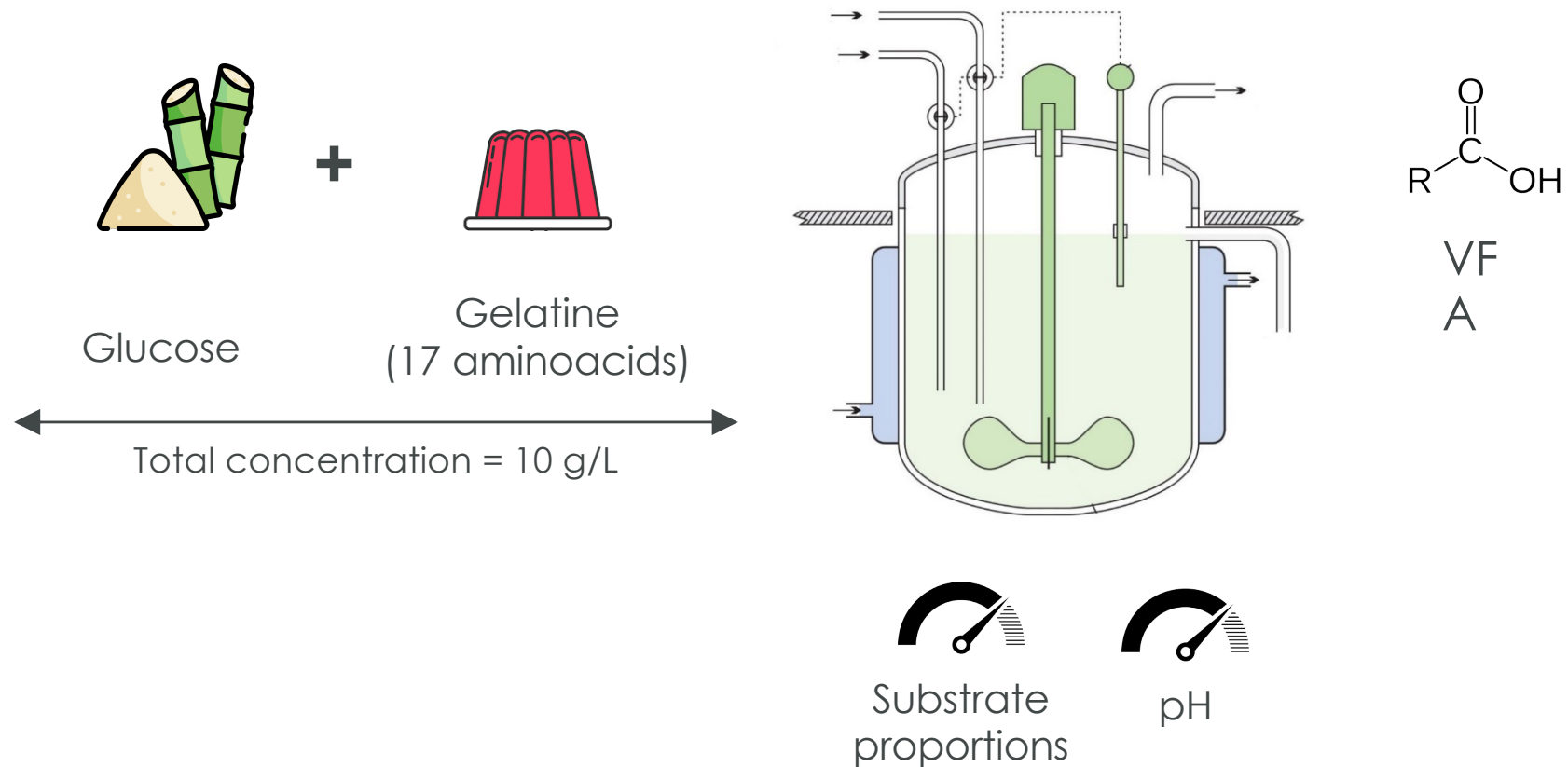


The model captures the tendencies observed with pH



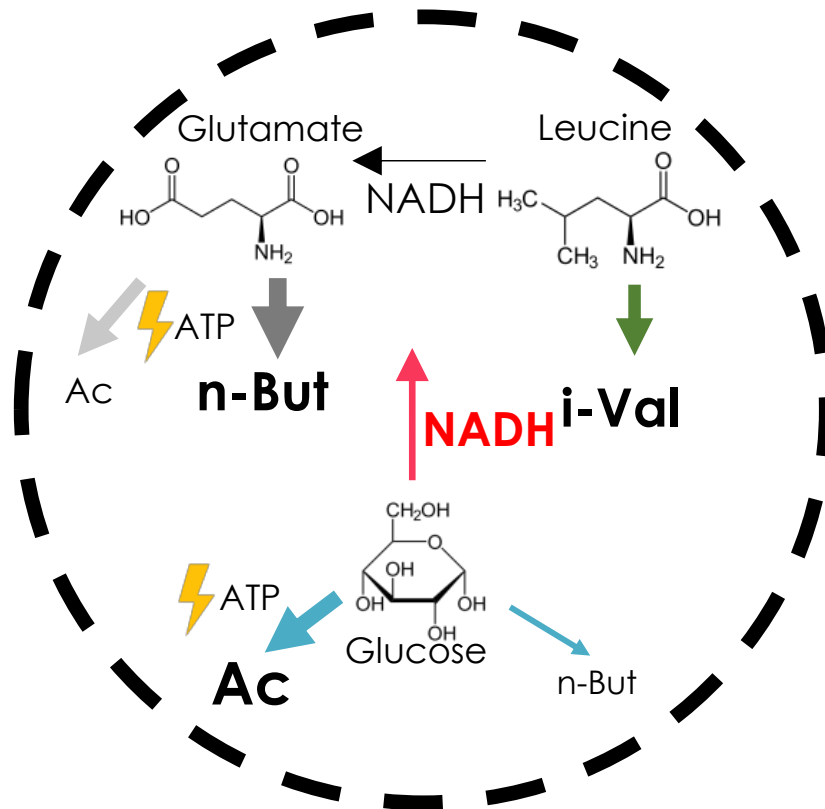


Modelling the cofermentation of gelatine and glucose





Cofermentation is not just adding two mono-fermentations



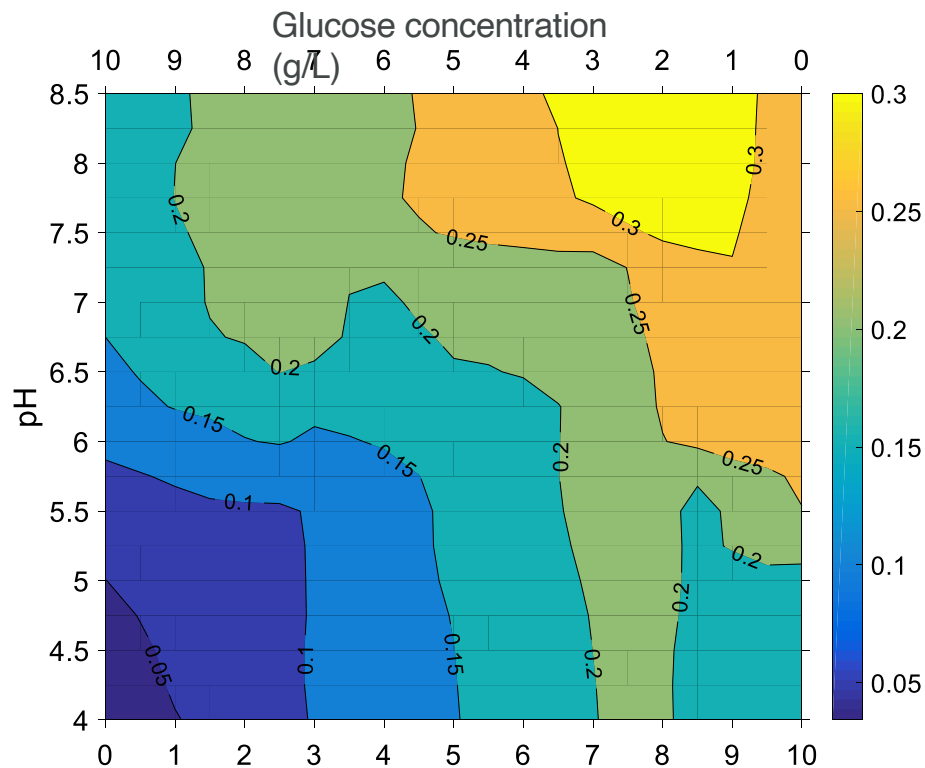
Cofermentation

$$\frac{\text{Max}(r_{\text{ATP,Prot}} + r_{\text{ATP,Glucose}})}{\sum r_{\text{NADH,Prot}} + \sum r_{\text{NADH,Glu}}} = 0$$

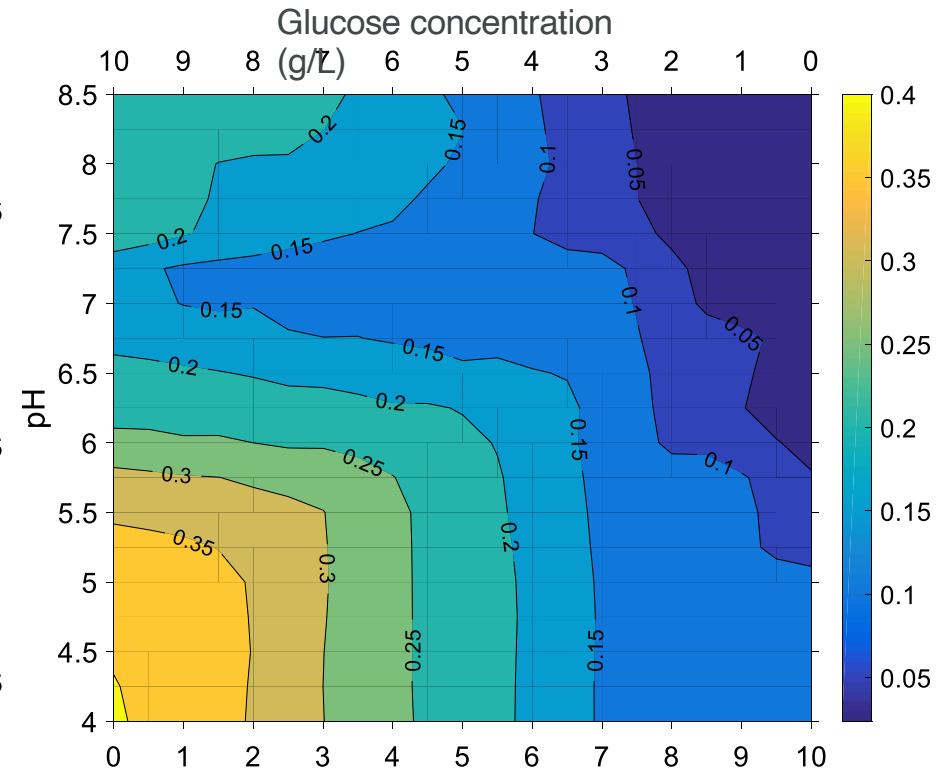


Changing operational conditions favour different VFA

Acetate yield (g/g_{Feeding})

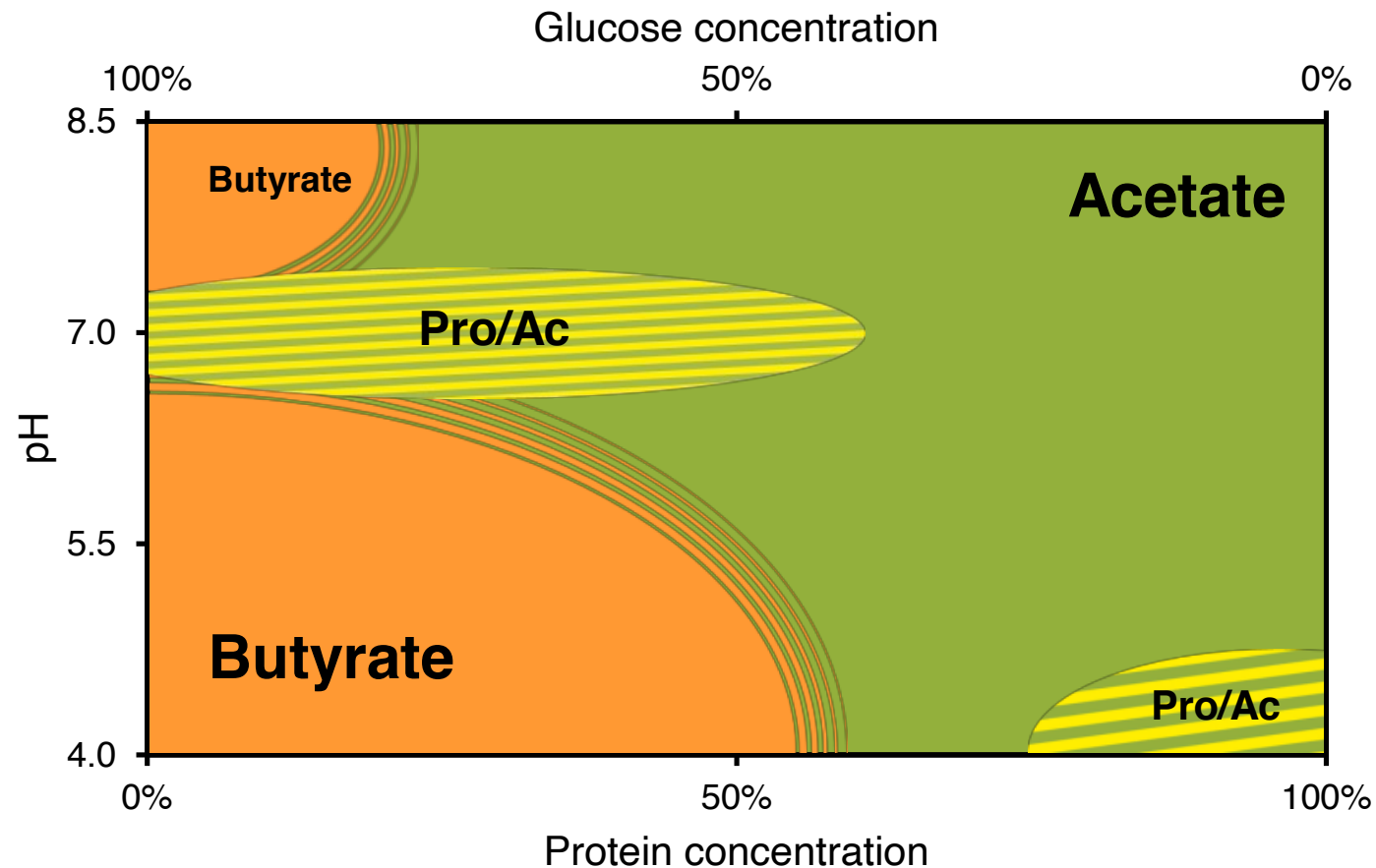


n-Butyrate yield (g/g_{Feeding})





The model can explore the operational space





Take home messages

- There are bioenergetic models available for **predicting the fermentation stoichiometry** of glucose, protein and their cofermentation.
- The **effect of main operational conditions** (e.g. pH or substrate proportions) on process selectivity is well captured by the model.
- These models have a direct use as **early-stage design tools** in the context of the carboxylate platform:
 - Select a (mixture of) waste streams to target a particular VFA(s)
 - Select the best pH for targeting a particular VFA(s)



Bioenergetic modelling for predicting the stoichiometry of anaerobic fermentation

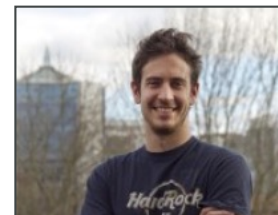


Alberte.regueiralopez@ugent.b

e



@Alberte Regueira



CMET

Center for Microbial Ecology and Technology



**GHENT
UNIVERSITY**



AGENDA AND HOUSEKEEPING

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Q&A Session Moderator: *Robbert*

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INFERRING METABOLIC INTERACTIONS FROM GENOME-SCALE DATA: ANAEROBIC DIGESTION ‘UNDER THE MICROSCOPE’

Adam Kovalovszki (adko@env.dtu.dk)

DTU Environment –
Technical University of Denmark



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Focus points

- ❑ What is genome-scale metabolic modeling (GSMM)?



Focus points

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- ❑ Who's who in the AD microbiome



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 - ↕
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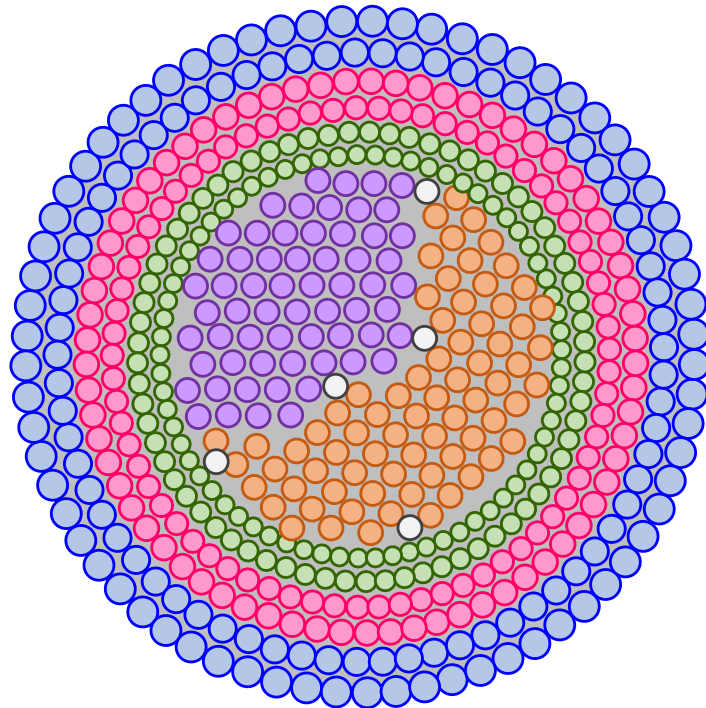


Focus points

- ❑ What is genome-scale metabolic modeling (GSMM)?
 - ❑ Who's who in the AD microbiome
 - ❑ Microbial interactions in AD



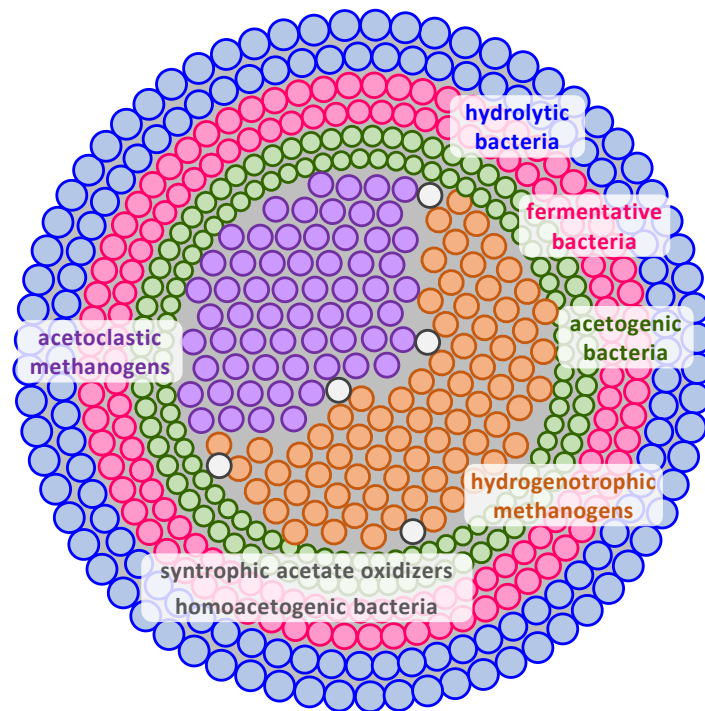
GSMM



METHANOGENIC
GRANULE
STRUCTURE
(theoretical)



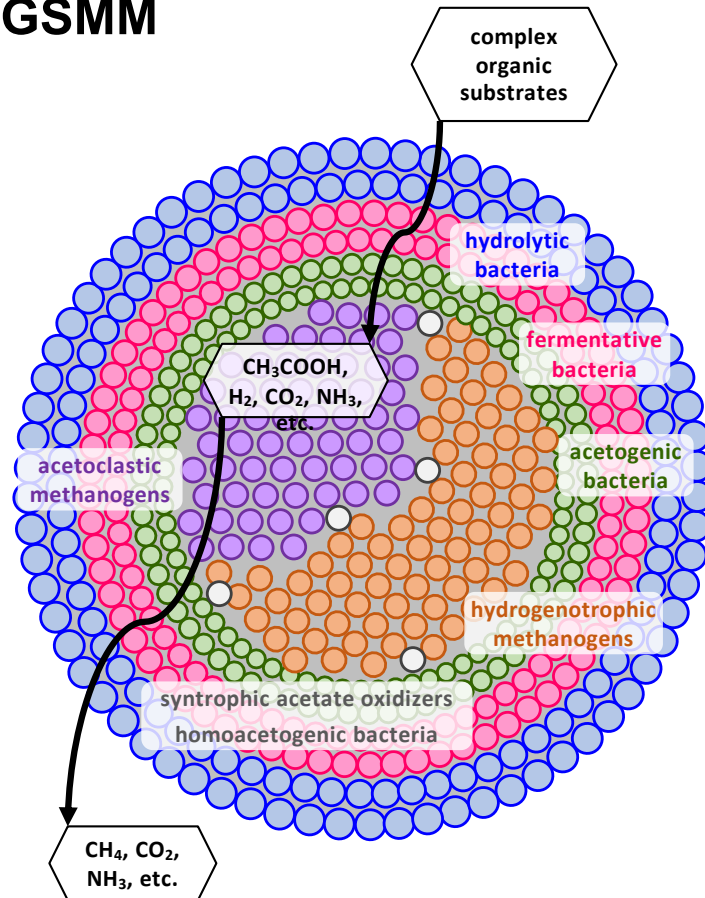
GSMM



**METHANOGENIC
GRANULE
STRUCTURE**
(theoretical)



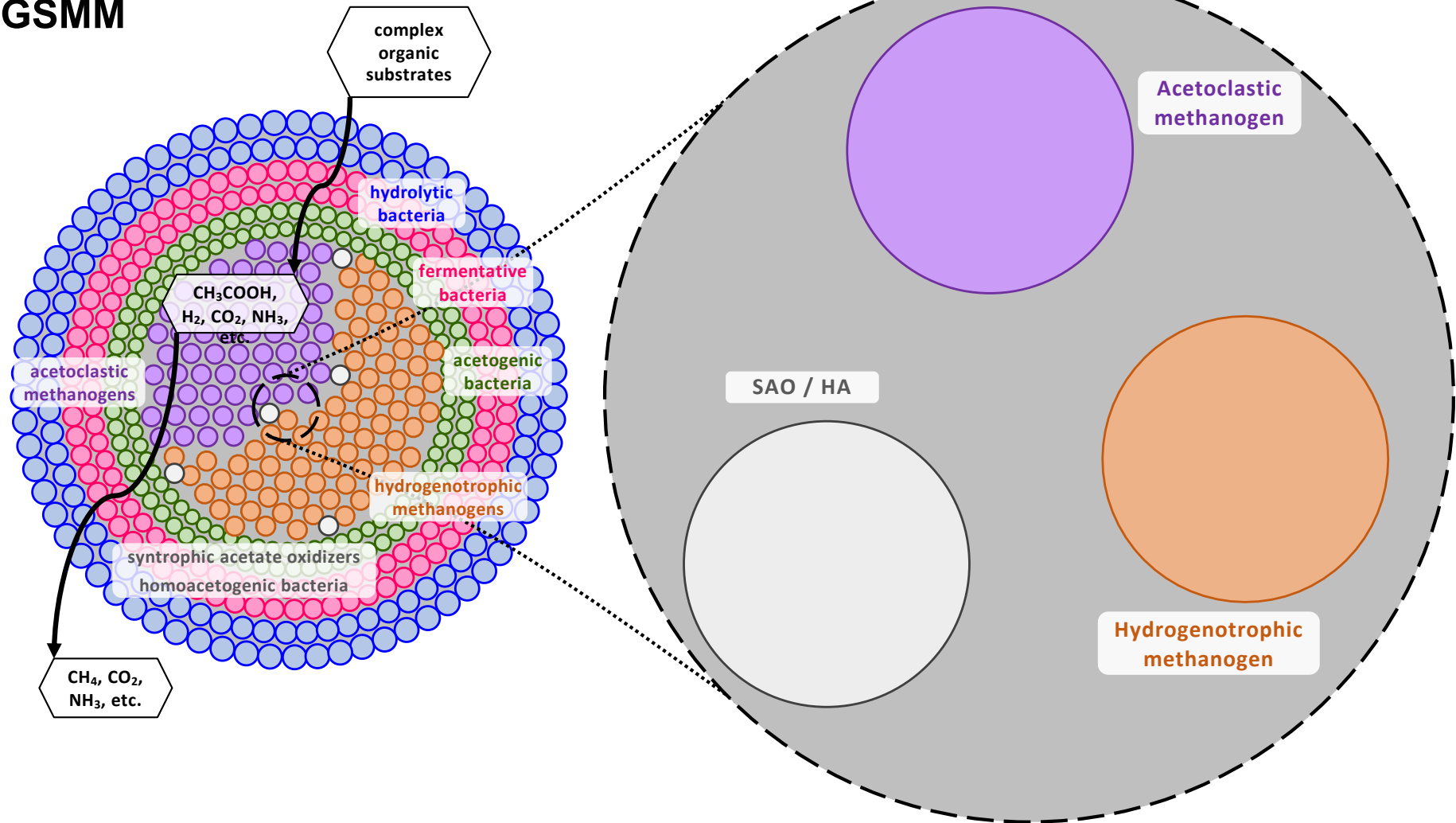
GSMM



METHANOGENIC
GRANULE
STRUCTURE
(theoretical)

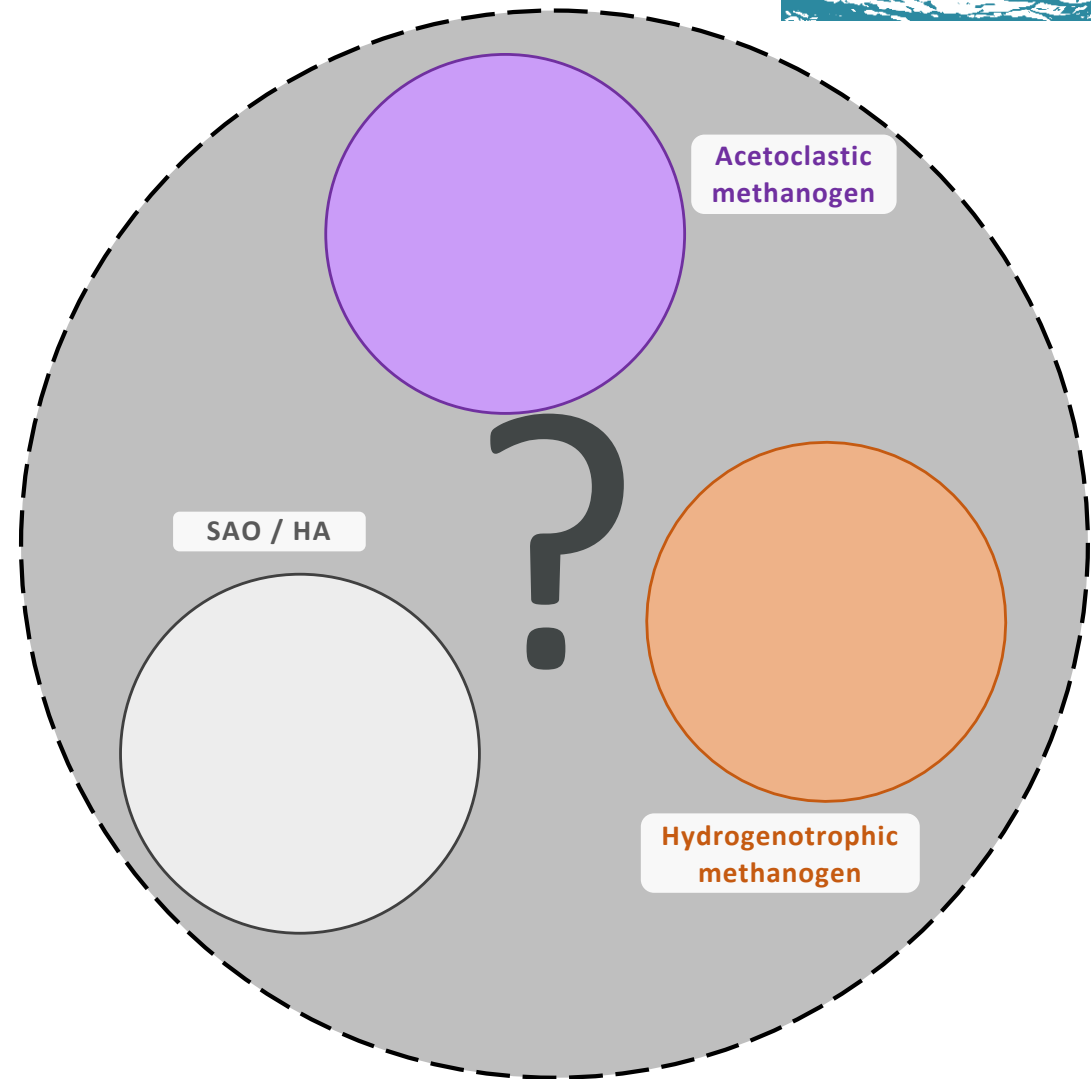


GSMM



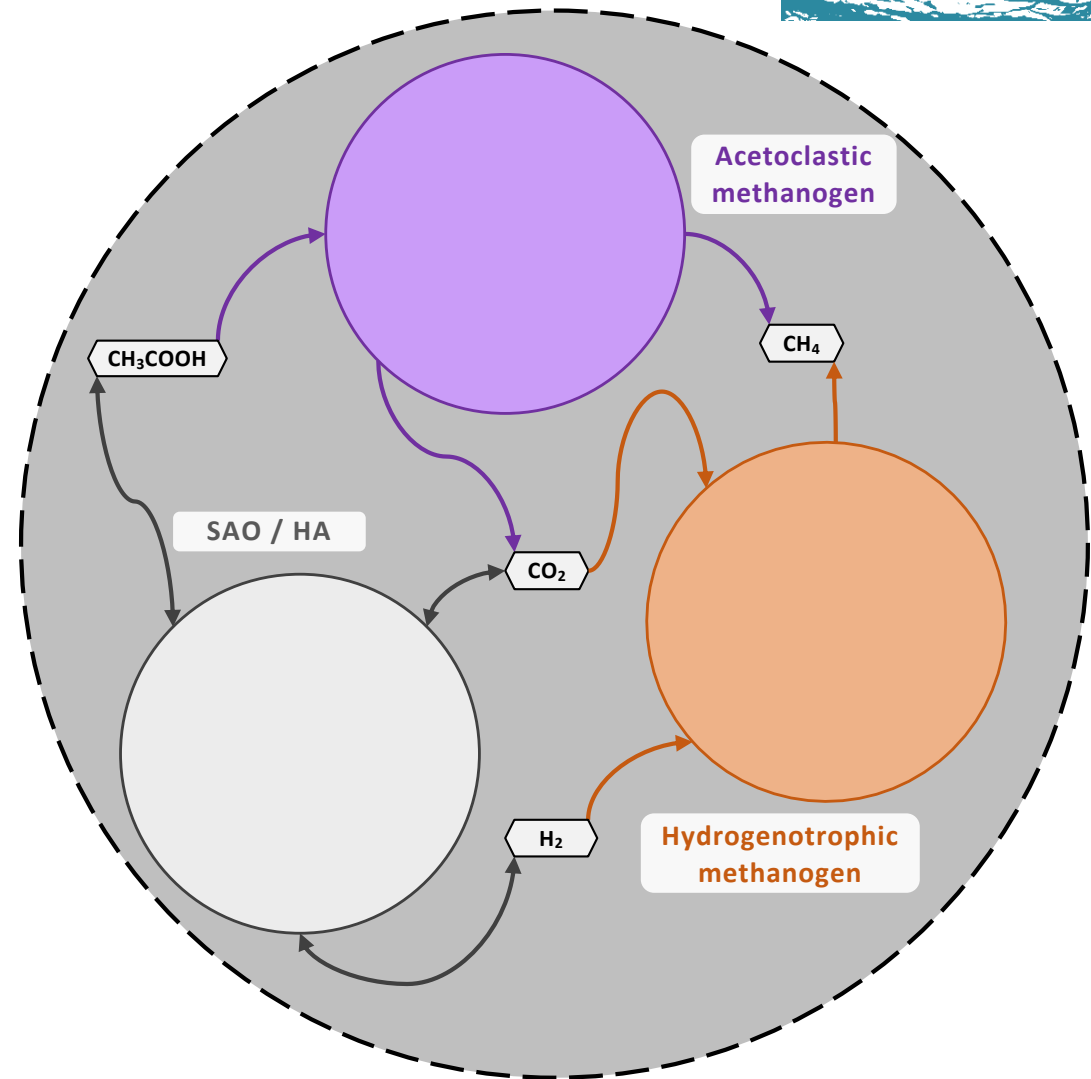


GSMM



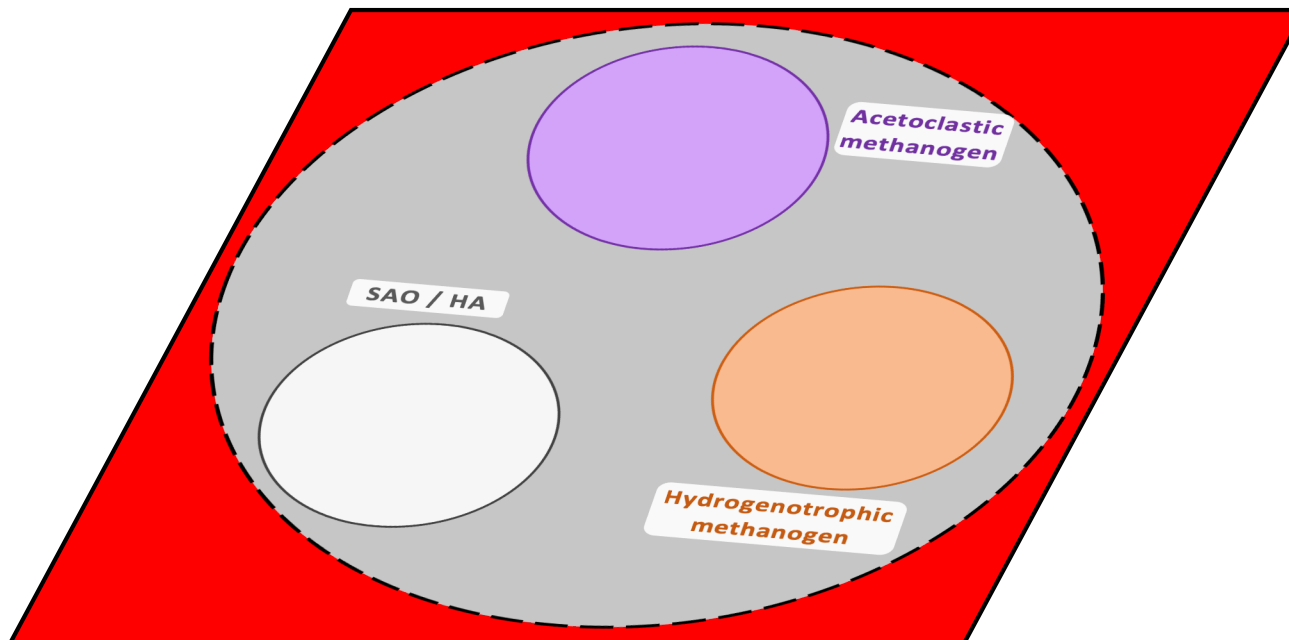


GSMM





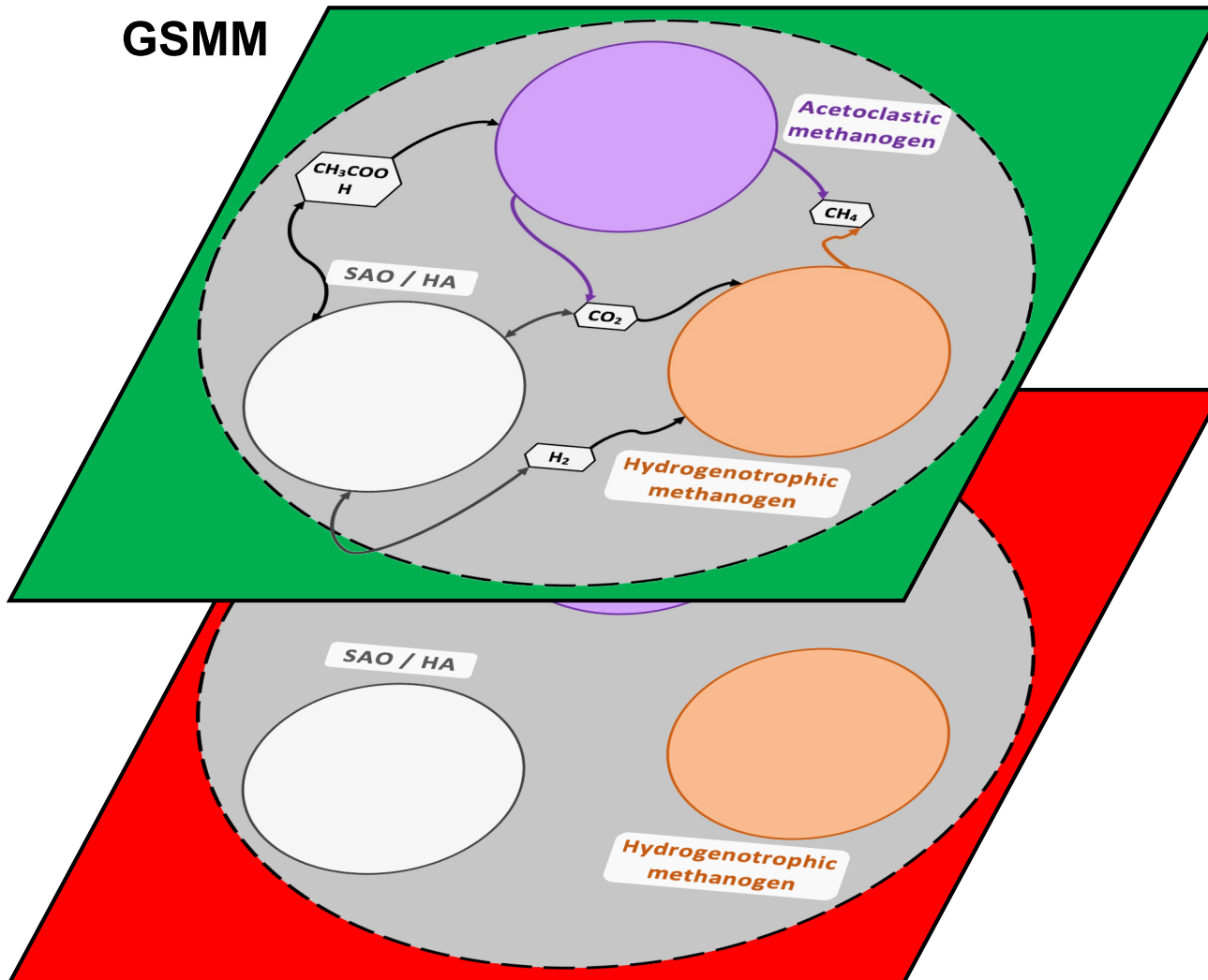
GSMM



Guilds (functions we assume)



GSMM

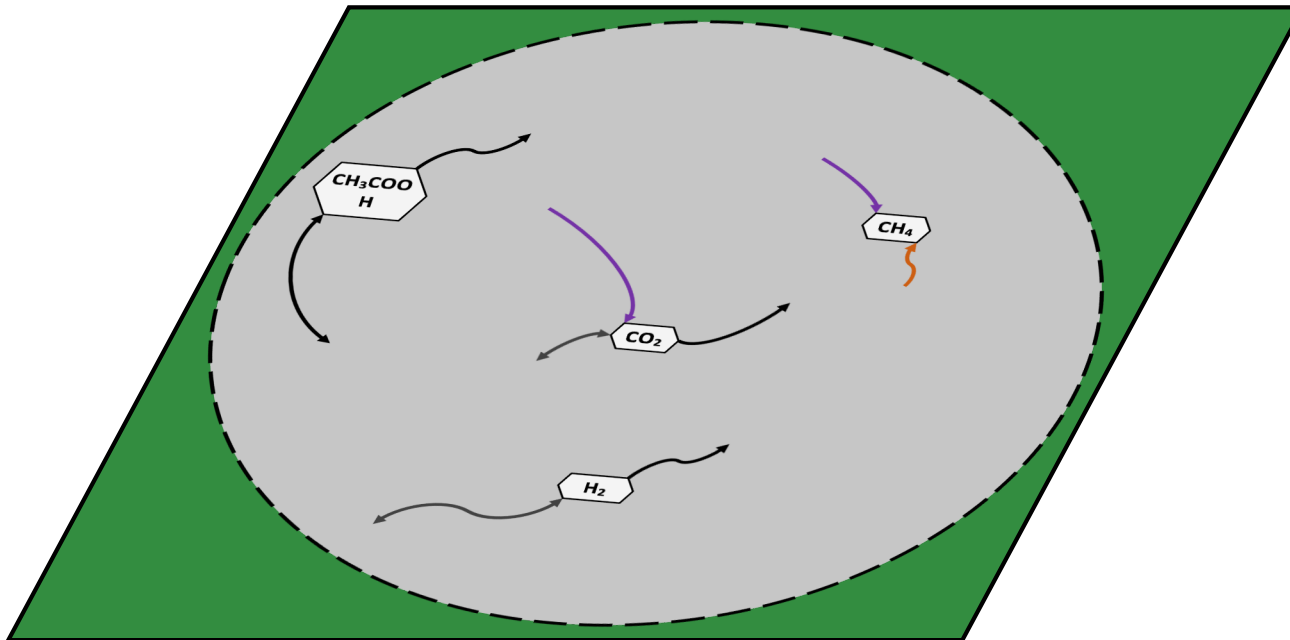


Phenotypes (what we "see")

Guilds (functions we assume)



GSMM



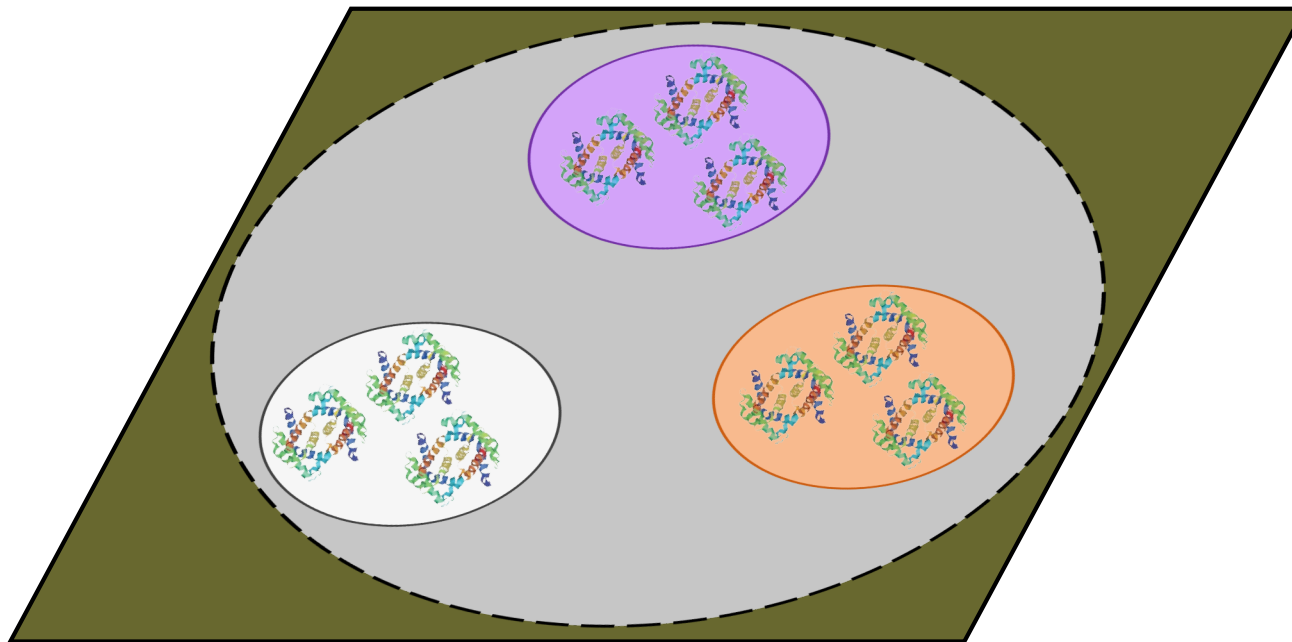
Phenotypes (*what we "see"*)

Metabolites → metabolomics

Guilds (functions we assume)



GSMM



Phenotypes (*what we "see"*)

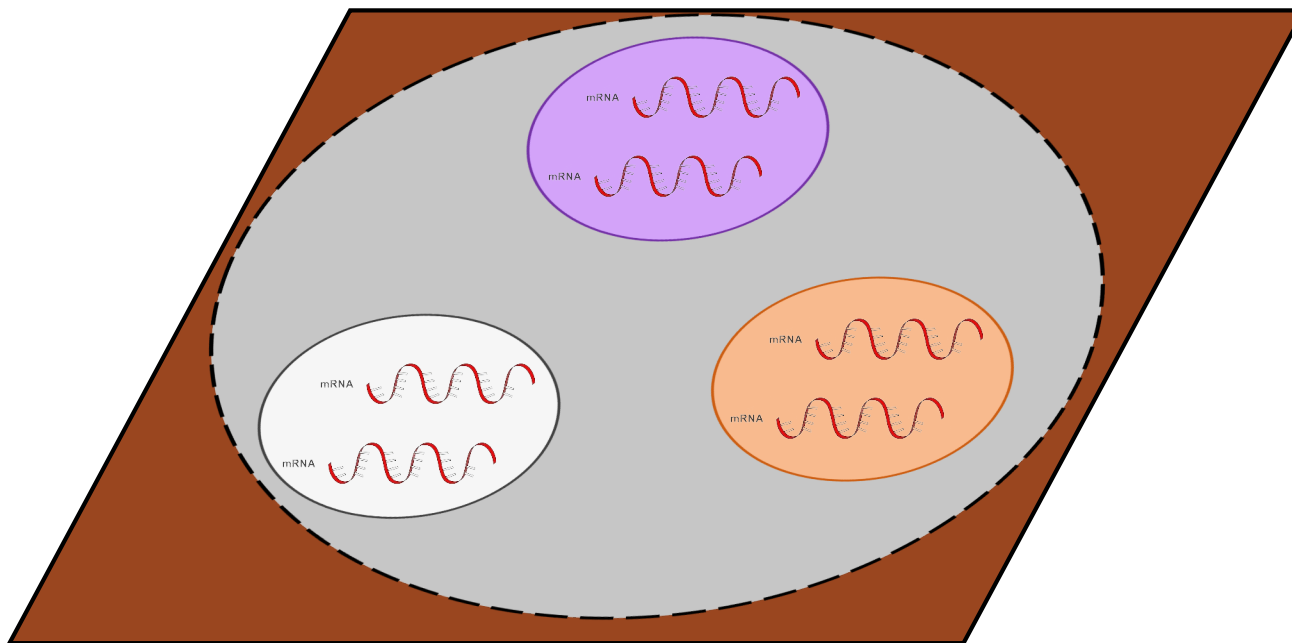
Metabolites → metabolomics

Proteins → Metaproteomics

Guilds (functions we assume)



GSMM



Phenotypes (*what we "see"*)

Metabolites → metabolomics

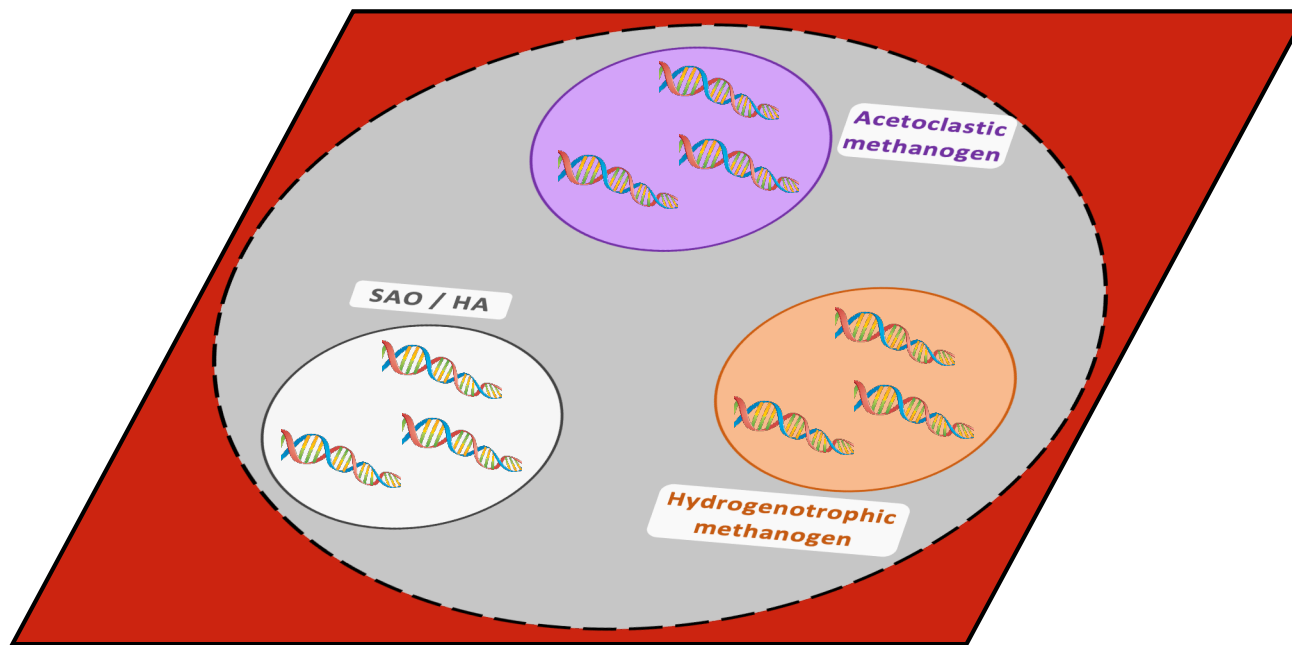
Proteins → Metaproteomics

mRNA → Metatranscriptomics

Guilds (functions we assume)



GSMM



Phenotypes (*what we "see"*)

Metabolites → metabolomics

Proteins → Metaproteomics

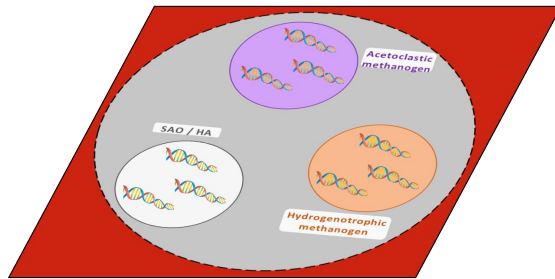
mRNA → Metatranscriptomics

DNA → Metagenomics
(16S → taxonomy)

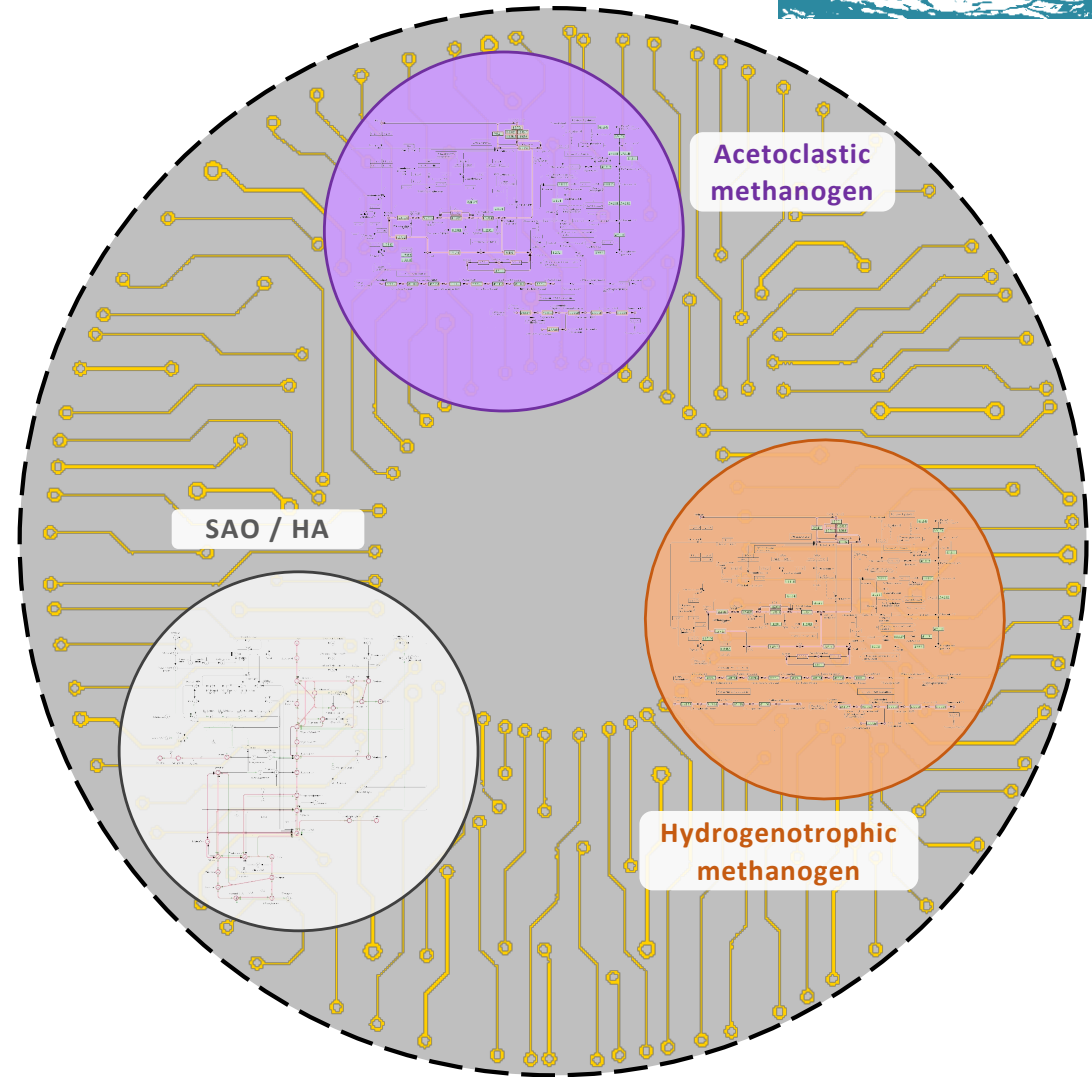
Guilds (functions we assume)



GSMM

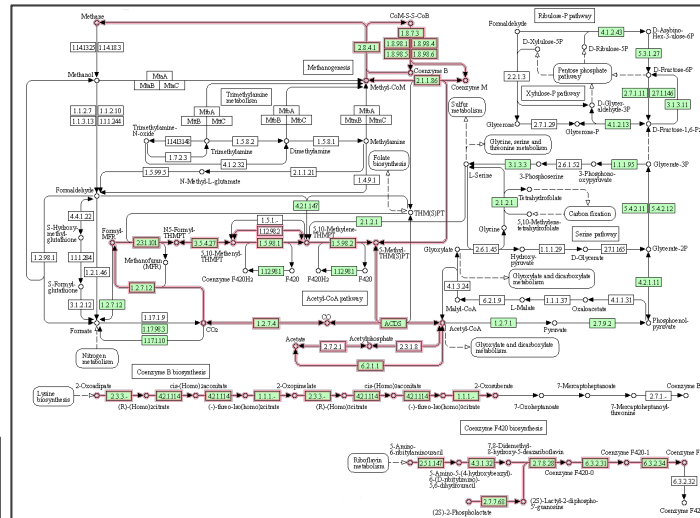
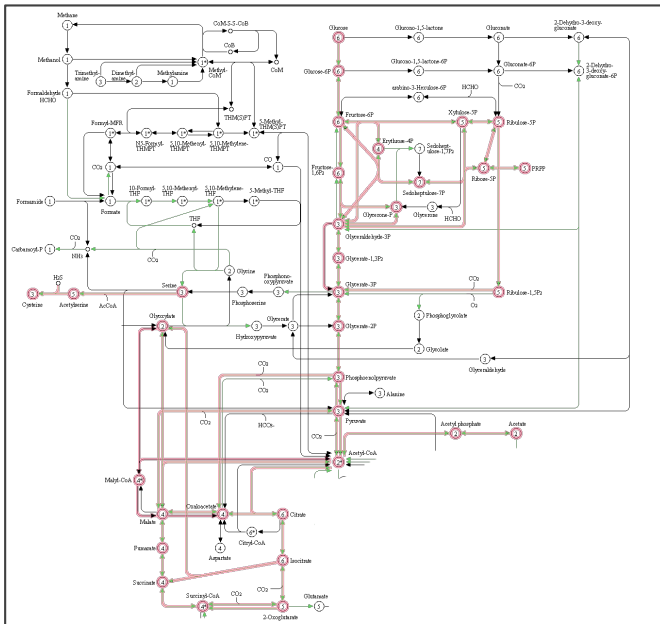


**METHANOGENIC
COMMUNITY
MODEL**
(in silico)



GSMM

SAO / HA

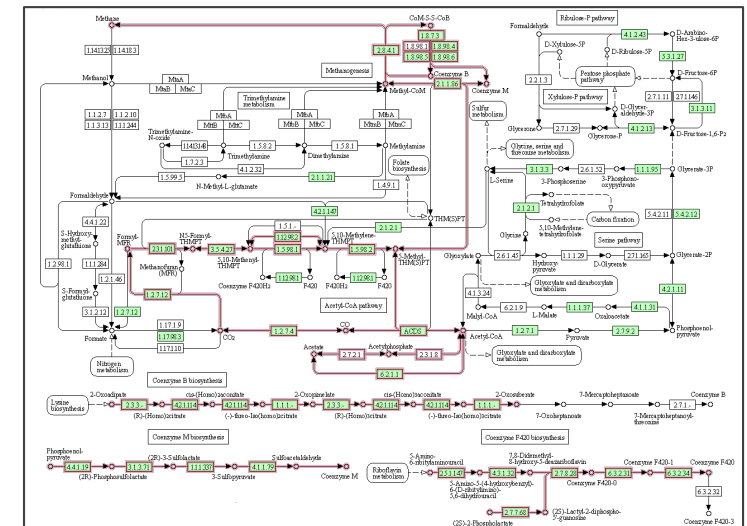


Acetoclastic methanogen



METHANOGENIC COMMUNITY MODEL (in silico)

Hydrogenotrophic methanogen



<https://www.kegg.jp/kegg/>

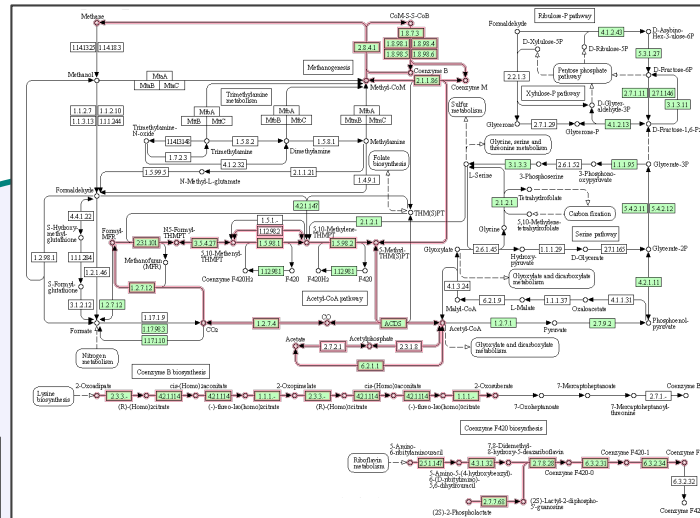
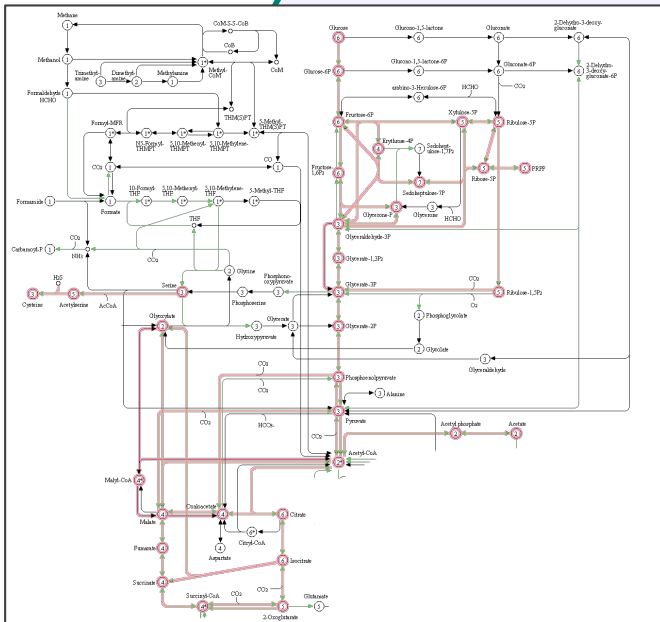
GSMM



Acetoclastic
methanogen

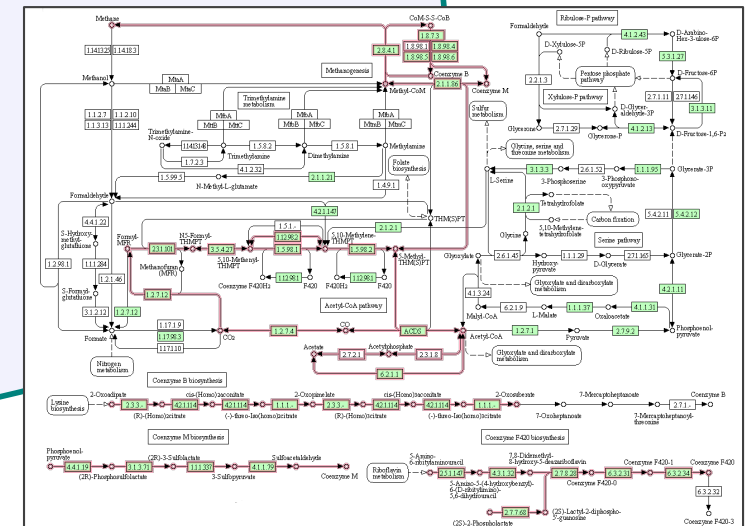
METHANOGENIC
COMMUNITY
MODEL
(in silico)

SAO / HA



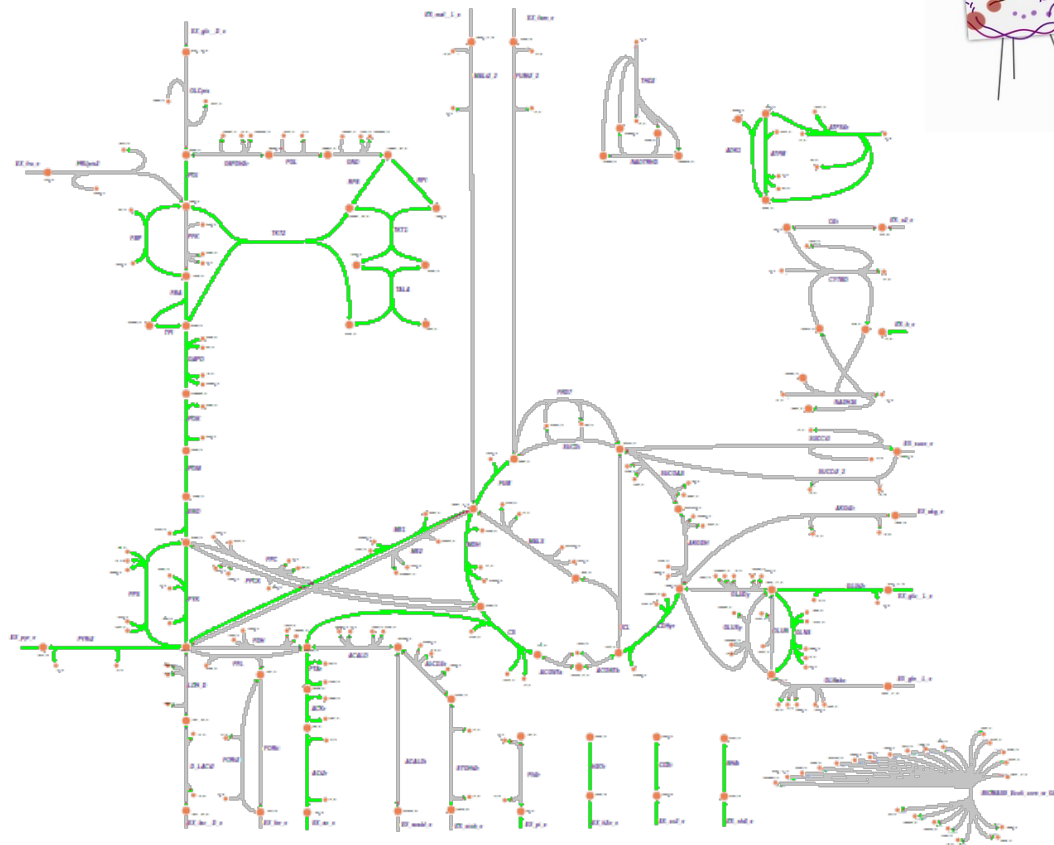
SHARED
MEDIUM

Hydrogenotrophic
methanogen





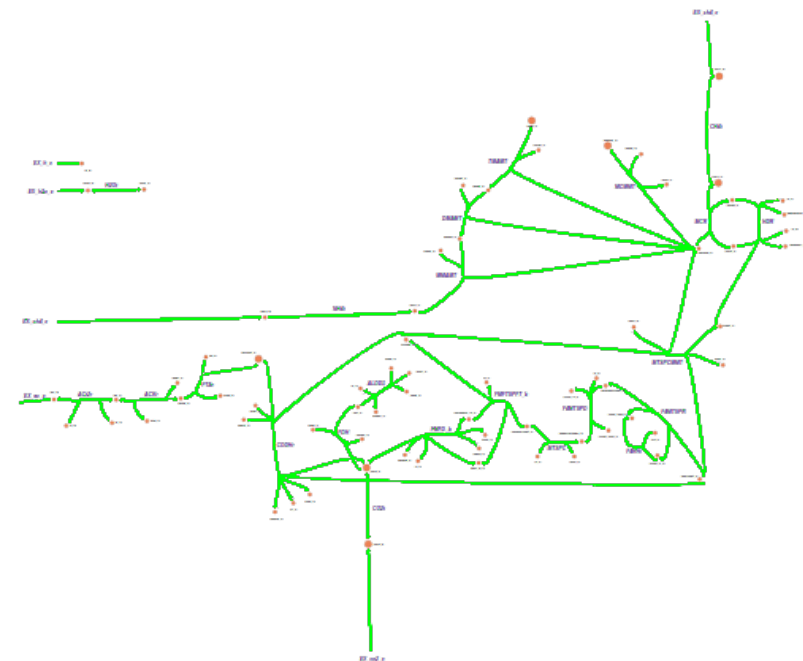
GSMM



Core carbon metabolism



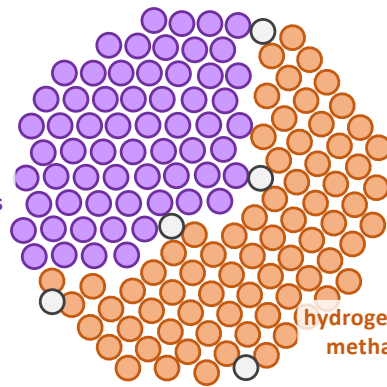
Methane metabolism





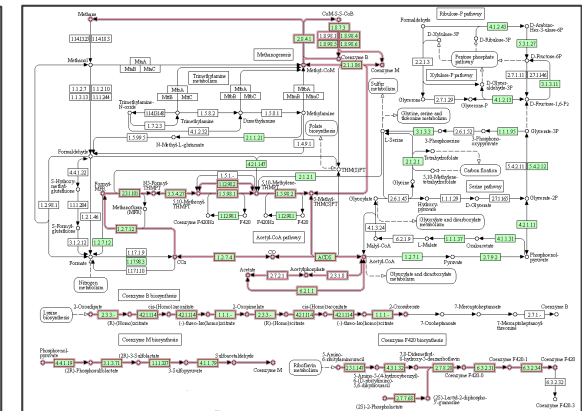
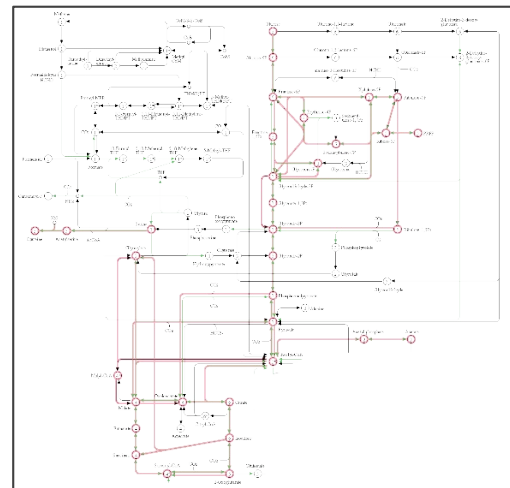
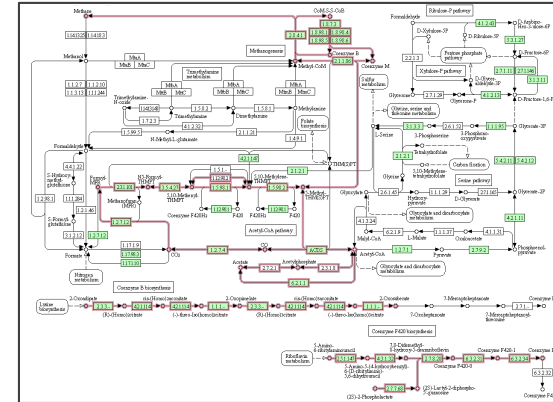
GSMM

acetoclastic
methanogens



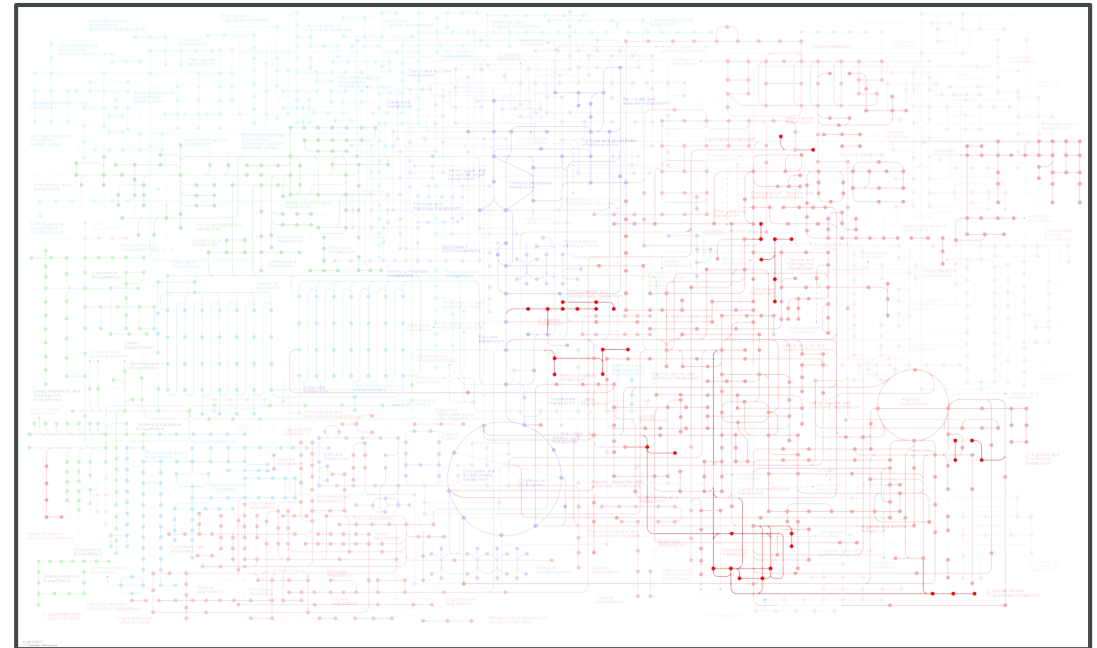
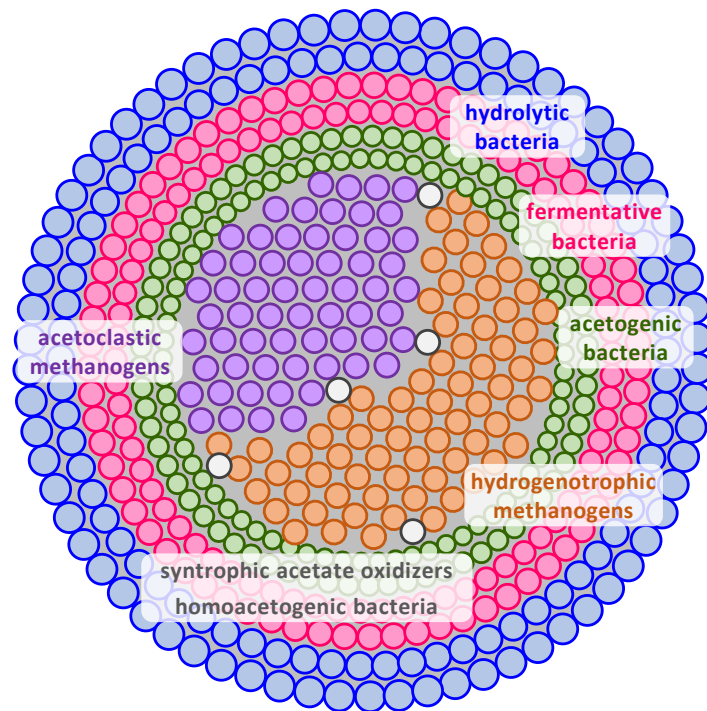
hydrogenotrophic
methanogens

syntrophic acetate oxidizers
homoacetogenic bacteria





GSMM



https://www.genome.jp/dbget-bin/www_bget?ko01100

Who's who in AD



Campanaro et al. *Biotechnol Biofuels* (2020) 13:25
https://doi.org/10.1186/s13068-020-01679-y

Biotechnology for Biofuels

RESEARCH

Open Access



New insights from the biogas microbiome by comprehensive genome-resolved metagenomics of nearly 1600 species originating from multiple anaerobic digesters

Stefano Campanaro^{1,3}, Laura Treu^{1,2*}, Luis M. Rodríguez-R⁴, Adam Kovalovszki², Ryan M. Ziels⁵, Irena Maus⁶, Xinyu Zhu², Panagiotis G. Kougias¹, Arianna Basile¹, Gang Luo⁶, Andreas Schlüter¹, Konstantinos T. Konstantinidis⁴ and Irini Angelidaki²

Abstract

Background: Microorganisms in biogas reactors are essential for degradation of organic matter and methane production. However, a comprehensive genome-centric comparison, including relevant metadata for each sample, is still needed to identify the globally distributed biogas community members and serve as a reliable repository.

Results: Here, 134 publicly available metagenomes derived from different biogas reactors were used to recover 1635 metagenome-assembled genomes (MAGs) representing different biogas bacterial and archaeal species. All genomes were estimated to be > 50% complete and nearly half ≥ 90% complete with ≤ 5% contamination. In most samples, specialized microbial communities were established, while only a few taxa were widespread among the different reactor systems. Metabolic reconstruction of the MAGs enabled the prediction of functional traits related to biomass degradation and methane production from waste biomass. An extensive evaluation of the replication index provided an estimation of the growth dynamics for microbes involved in different steps of the food chain.

Conclusions: The outcome of this study highlights a high flexibility of the biogas microbiome, allowing it to modify its composition and to adapt to the environmental conditions, including temperatures and a wide range of substrates. Our findings enhance our mechanistic understanding of the AD microbiome and substantially extend the existing repository of genomes. The established database represents a relevant resource for future studies related to this engineered ecosystem.

Keywords: Anaerobic digestion, Metagenome-assembled genomes, Biogas, Microbial community structure, Functional reconstruction

Background

Anaerobic environments are ubiquitous in the biosphere. Some examples are the digestive tract of animals, paddy fields, wetlands and aquatic sediments. These

environments play crucial roles in the degradation of organic matter and in the global carbon cycle. The anaerobic digestion (AD) process has great societal importance since it reduces our dependence on fossil fuels via its ability to generate methane within engineered bioreactors [1]. For these reasons, the AD process has been widely established as an efficient metabolic route allowing the conversion of organic wastes, agricultural residues and renewable primary products into energy and

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Full list of author information is available at the end of the article

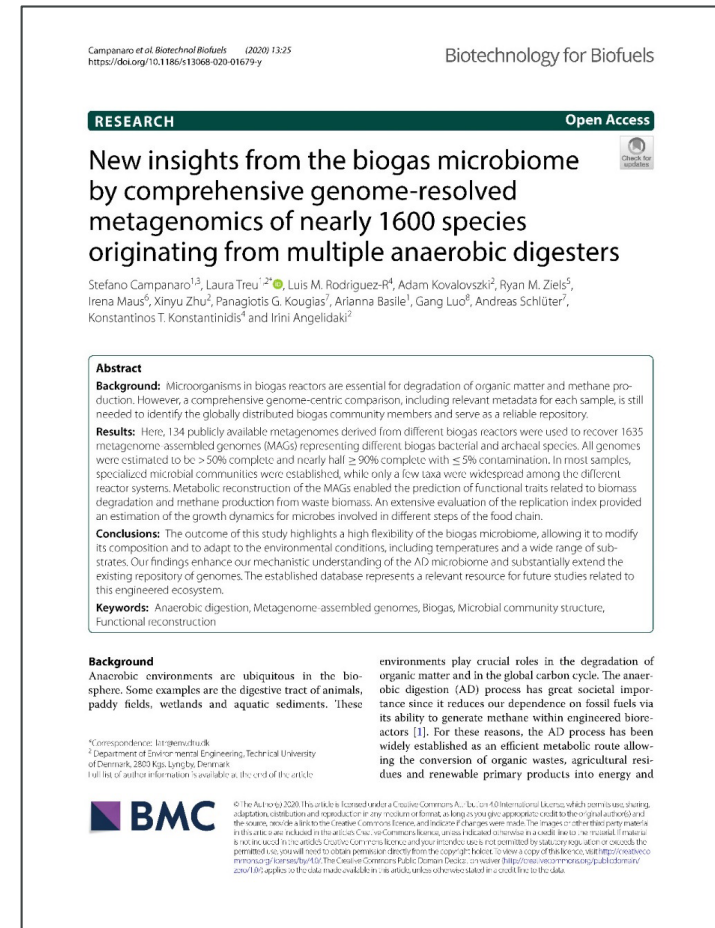


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Who's who in AD

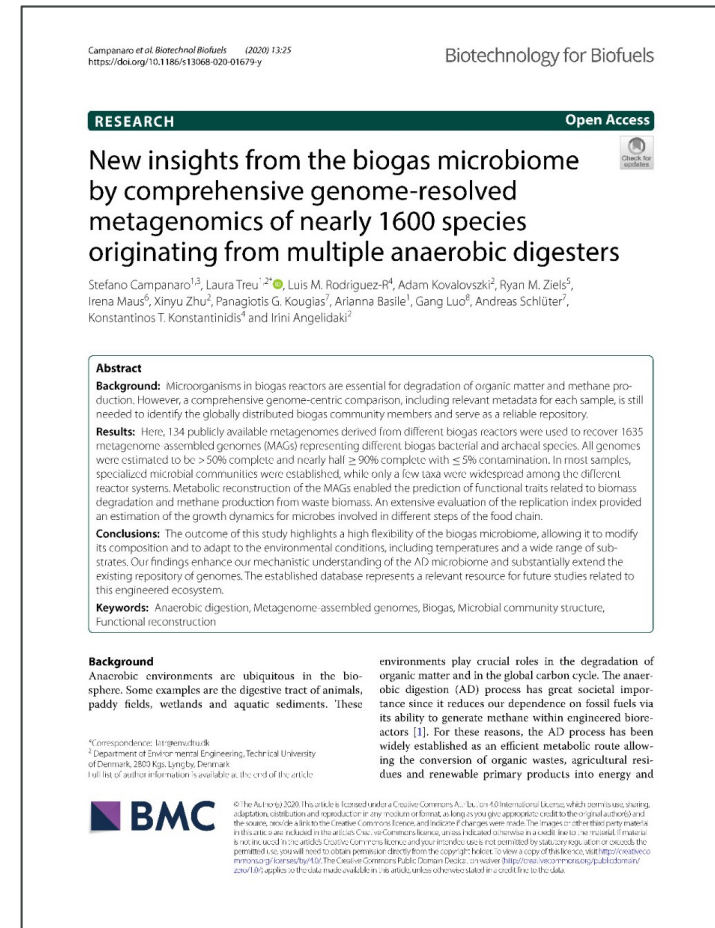
- 18 studies/sources





Who's who in AD

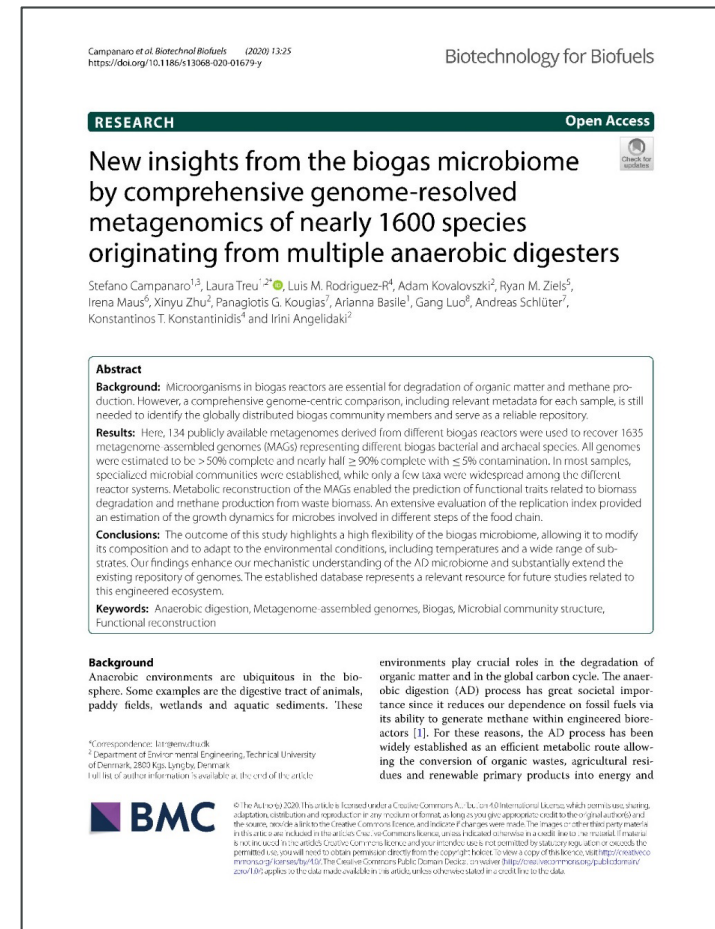
- 18 studies/sources
- 134 metagenomes





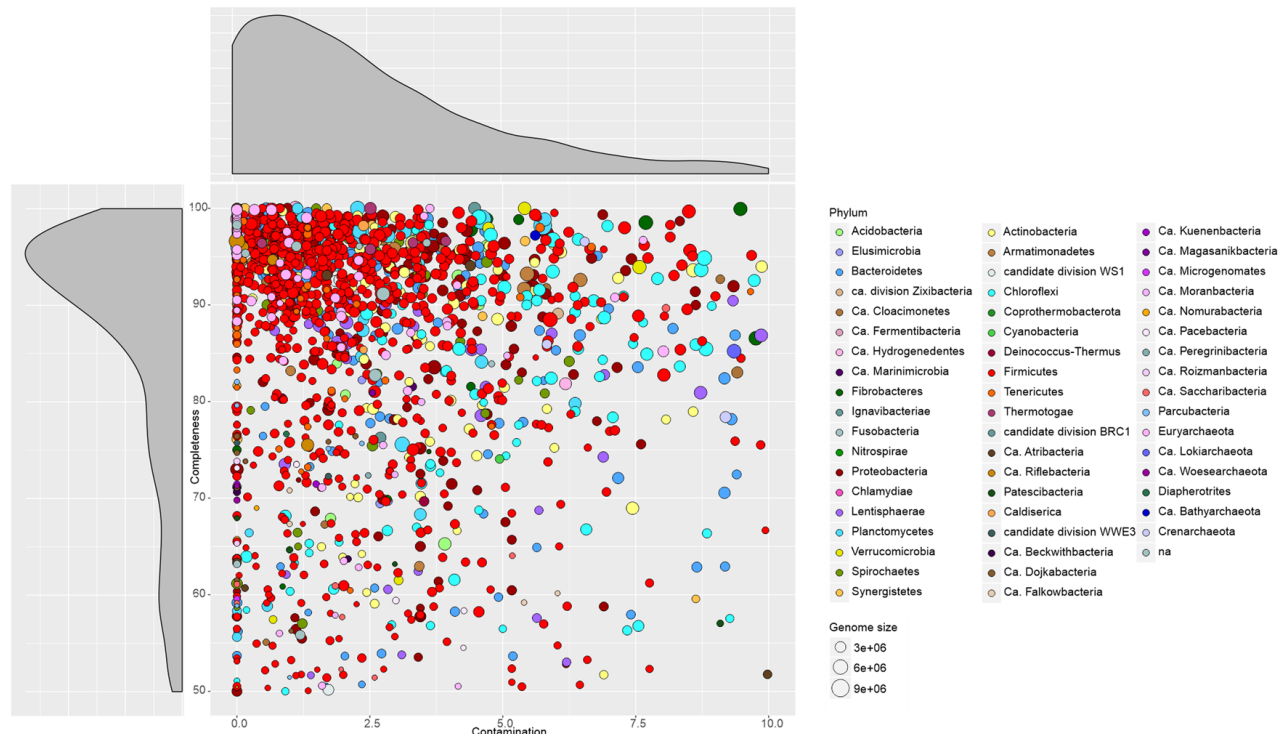
Who's who in AD

- 18 studies/sources
- 134 metagenomes
- 1635 metagenome-assembled genomes (MAGs)





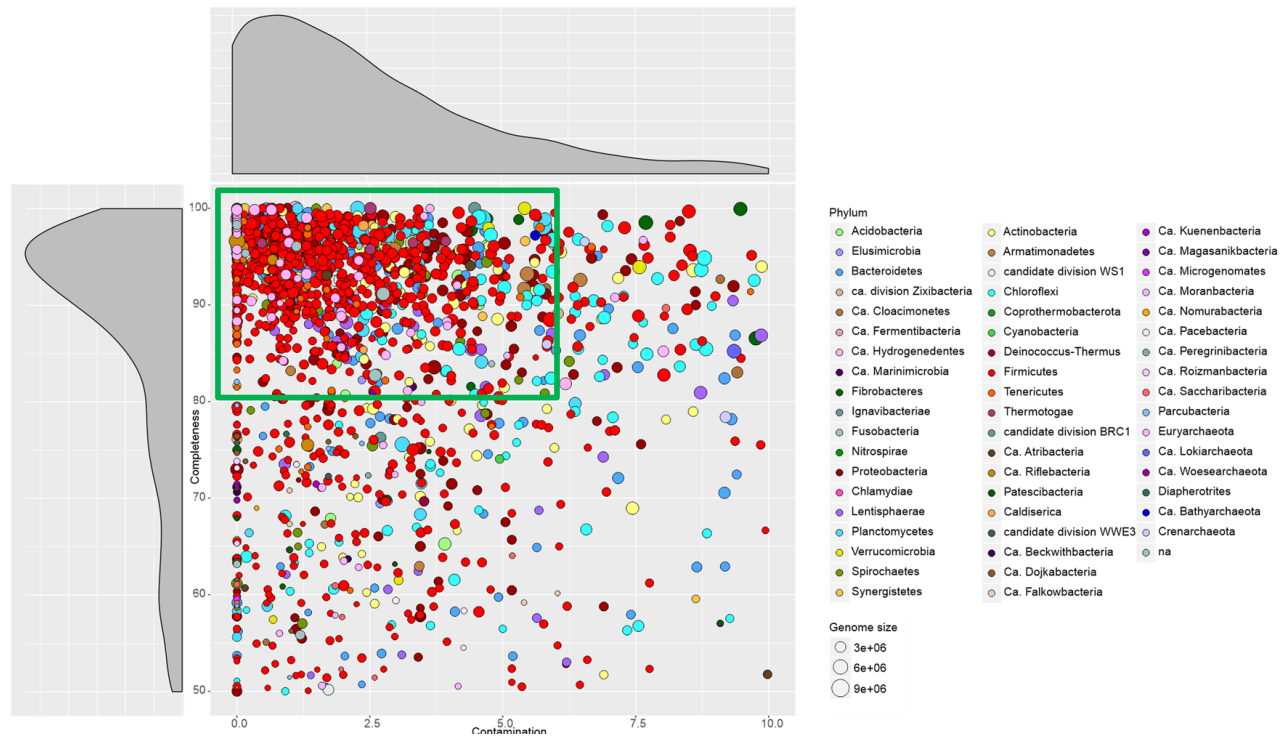
Who's who in AD



Campanaro *et al.*, 2020, New insights ...



Who's who in AD



Majority of MAGs with

- High level of completeness
- Low contamination

Microbial interactions in AD



Original Research Article

Revealing metabolic mechanisms of interaction in the anaerobic digestion microbiome by flux balance analysis

Arianna Basile^a, Stefano Campanaro^{a,b,*}, Adam Kovalovszki^c, Guido Zampieri^{a,d},
Alessandro Rossi^a, Irini Angelidaki^e, Giorgio Valle^{a,1}, Laura Treu^{a,1}

^a Department of Biology, University of Padova, Via U. Martini 260/5, 35121, Padova, Italy

^b CIRM Biotechnology Center, University of Padova, 35131, Padova, Italy

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ARTICLE INFO

Keywords:

Anaerobic digestion/flux balance analysis/
genome scale metabolic models/microbial in-
teractions/renewable energy

ABSTRACT

Anaerobic digestion is a key biological process for renewable energy, yet the mechanistic knowledge on its hidden microbial dynamics is still limited. The present work charted the interaction network in the anaerobic digestion microbiome via the full characterization of pairwise interactions and the associated metabolite ex- changes. To this goal, a novel collection of 836 genome-scale metabolic models was built to represent the functional capabilities of bacteria and archaea species derived from genome-centric metagenomics. Dominant microbes were shown to prefer mutualistic, parasitic and commensalistic interactions over neutralism, amensalism and competition, and are more likely to behave as metabolite importers and profitters of the coexistence. Additionally, external hydrogen injection positively influences anaerobic dynamics by promoting commensalism over amensalism. Finally, exchanges of glucogenic amino acids were shown to overcome auxotrophies caused by an incomplete triarabonyl acid cycle. Our novel strategy predominate the most favorable growth conditions for the microbes, overall suggesting strategies to increase the biogas production efficiency. In principle, this approach could also be applied to microbial populations of biomedical importance, such as the gut microbiome, to allow a broad inspection of the microbial interplays.

1. Introduction

Microorganisms play an important role in all fields of biological relevance, ranging from human health (Clemente et al., 2012) to biotechnology (Lehman et al., 2015). In particular, diverse microbiomes may have various responsibilities, from causing diseases to influencing applied processes (e.g. biogas production) (Zhu et al., 2020) or the synthesis of polymeric substances (Chow et al., 2008). Microbial net- works, however, are still poorly understood due to difficulties in isolating most of the microbial species and to the heterogeneous nature of their interactions (Keller et al., 2018). While the inspection of a core microbiome might reveal which species are the key players for a specific process (Faith, 2015), the role of rare members still remains to be clarified (Poulsen et al., 2017). Microbial cooperation is extremely important for environmental niche colonization and completing com- plex activities (Stolyar et al., 2007), which single species could not

perform independently (Day et al., 2004). This is the case in anaerobic digestion (AD), which is a biotechnological process that produces a potent renewable energy carrier called biogas (Ventelakis and Goulet, 2017). During biogas production, when acetoclastic methanogenic archaea are inhibited, a pivotal role is played by hydrogenotrophic methanogenic archaea and syntrophic acetate oxidizing bacteria (Moo- bak et al., 2016) (SAOB). An example is the mutualism between the hydrogen-utilizing methanogen *Methanococcus bozemans* and the SAOB *Syntrophomonas schinkii*, *[Clostridium] ultunense*, and *Veillonella* *acetoxydans* (Vierstraete et al., 2019). SAOB oxidize acetate to formate or to H₂ and carbon dioxide (CO₂). The bacteria rely on archaeal activity, because acetate oxidation rapidly becomes endergonic when H₂ accumulates (Stams and Plugge, 2009). Indeed, subsequently H₂-utiliz- ing methanogens convert these substrates to methane (CH₄) (Treu et al., 2019).

Although direct microbial cultivation and phenotyping experiments

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¹ equal contribution.

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Microbial interactions in AD

- 836 GSMMs





Microbial interactions in AD

- 836 GSMMs
- 349k+ pair-wise interactions studied





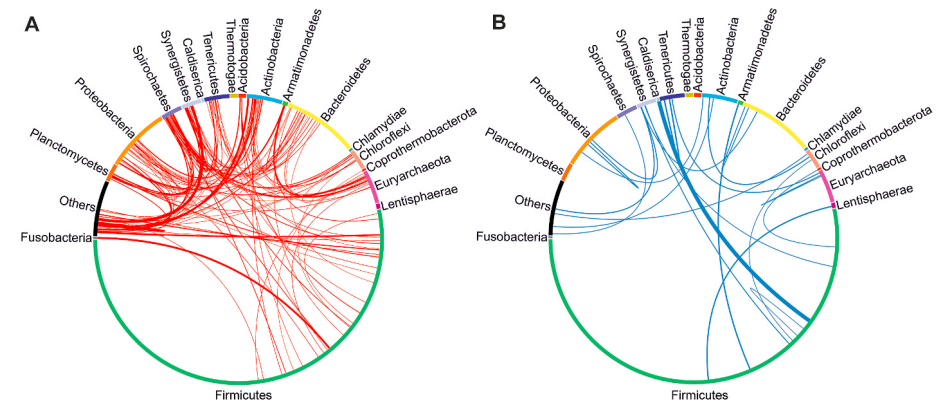
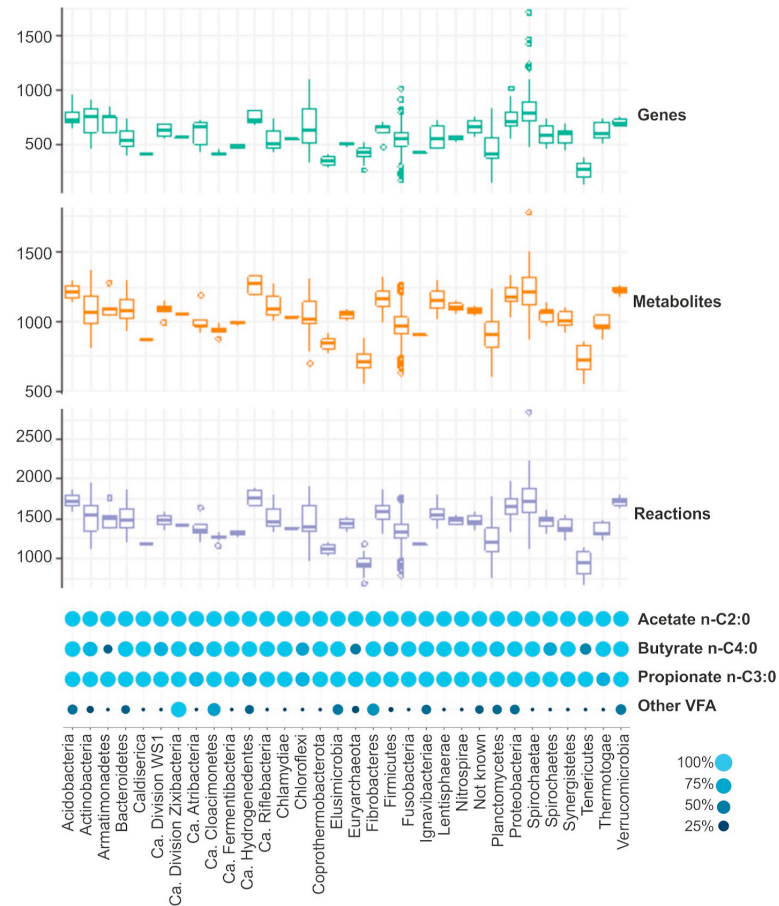
Microbial interactions in AD

- 836 GSMMs
- 349k+ pair-wise interactions studied
- Dominant species
 - Mutualistic (+/+)
 - Parasitic (+/-)
 - Commensalistic (+/0)



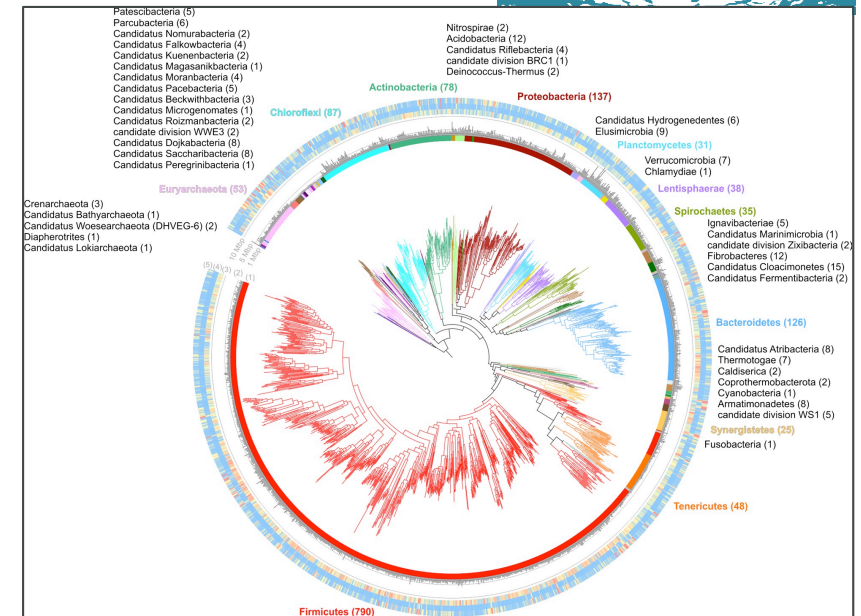
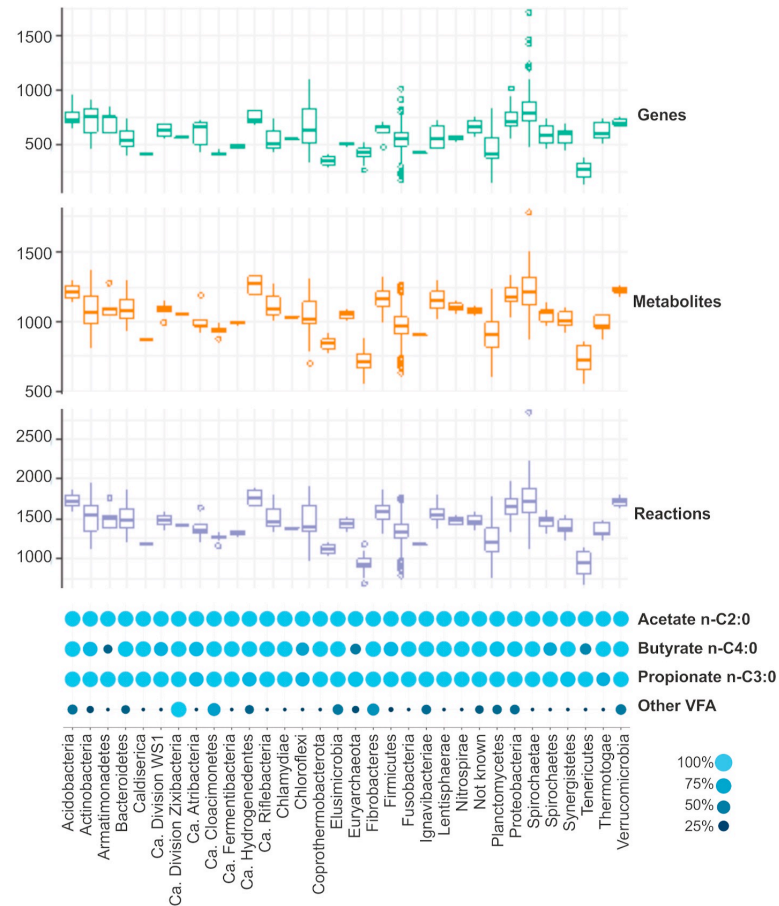


Microbial interactions in AD

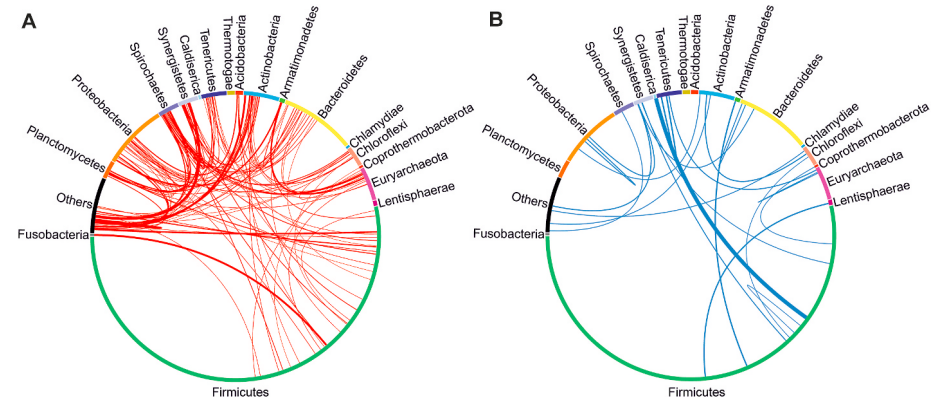




Microbial interactions in AD



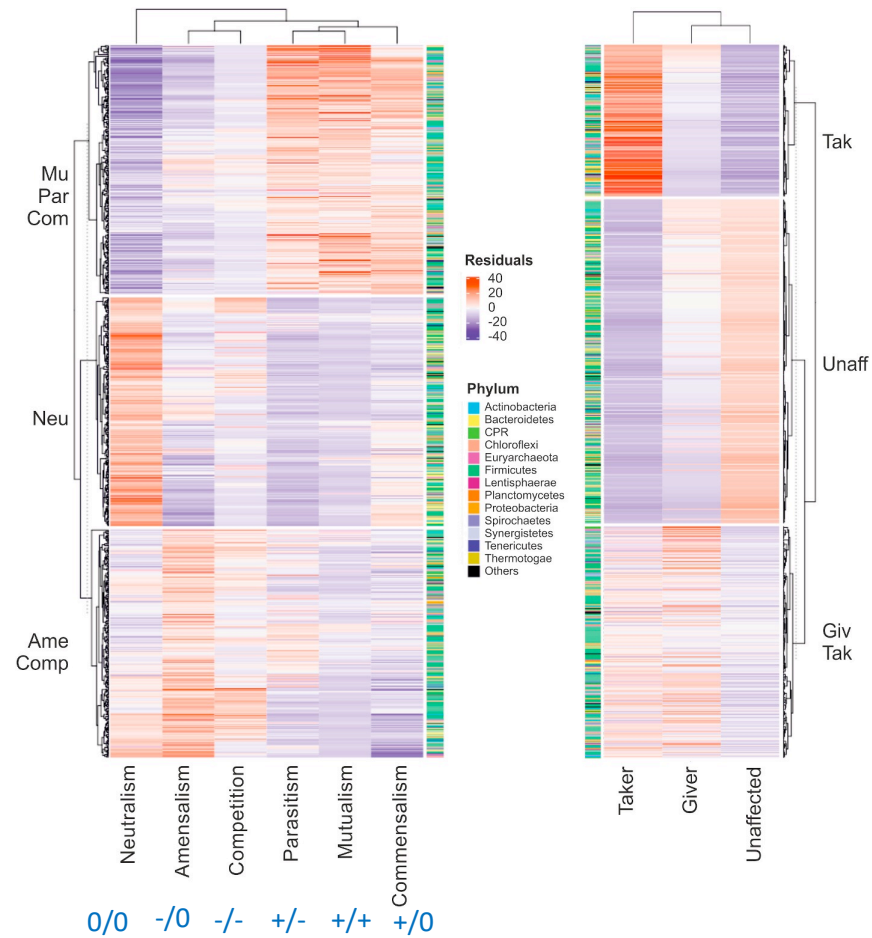
Campanaro *et al.*, 2020, New insights ...



Basile *et al.*, 2020, Revealing metabolic ...



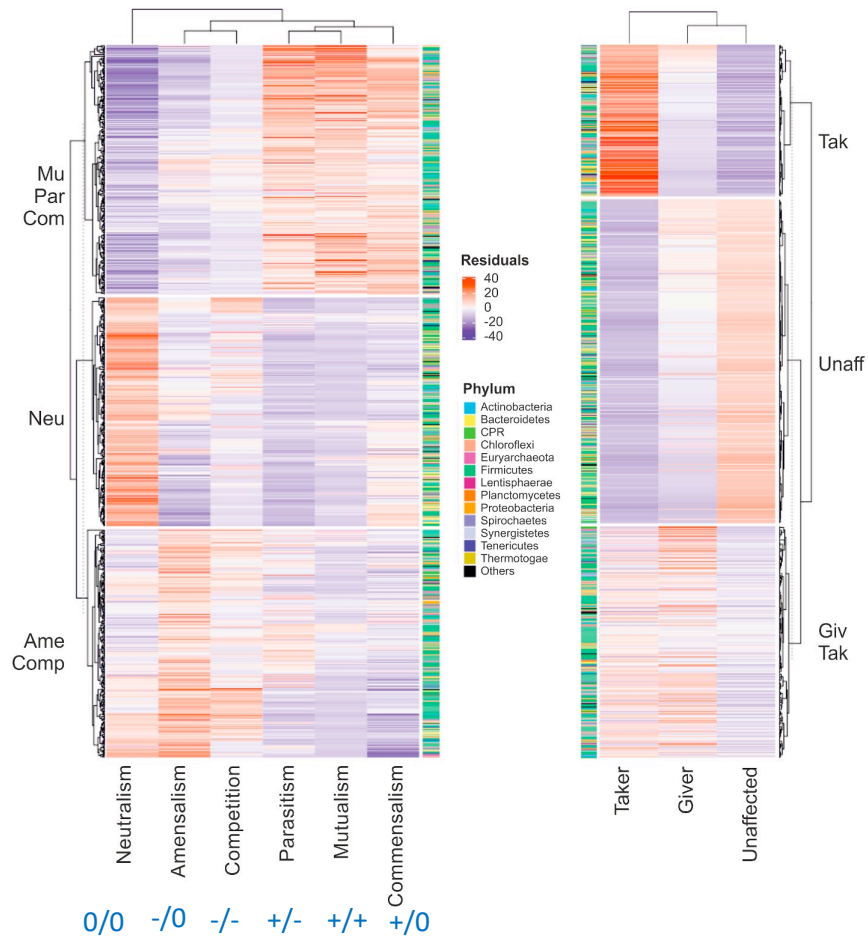
Microbial interactions in AD



Basile *et al.*, 2020, Revealing metabolic ...

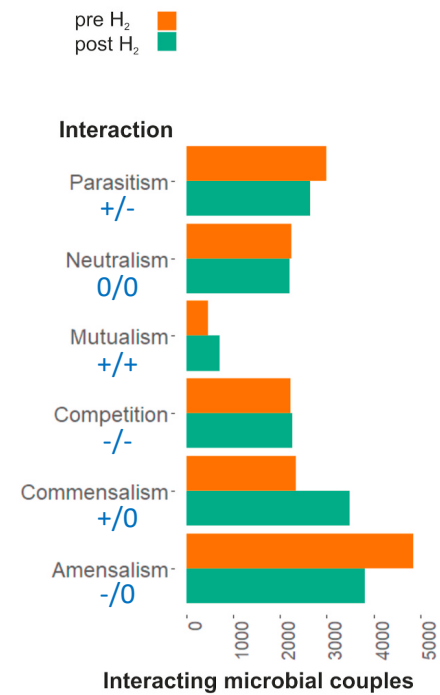


Microbial interactions in AD

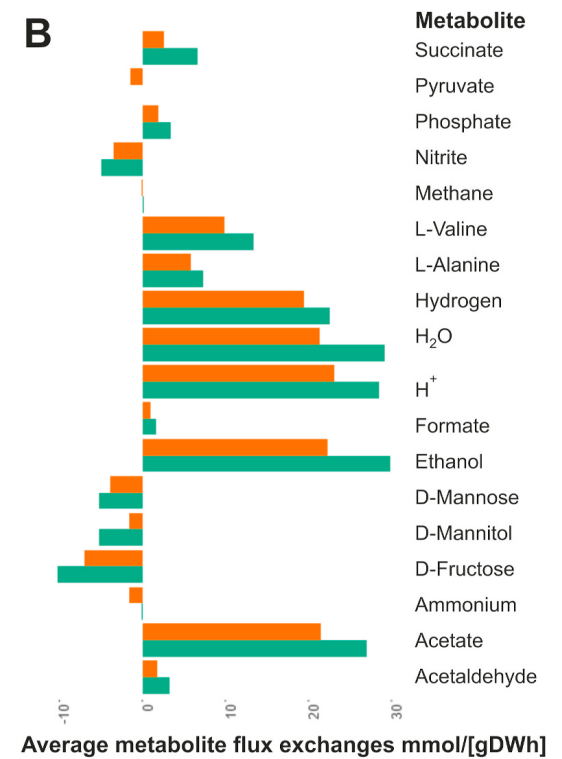


Basile *et al.*, 2020, Revealing metabolic ...

A



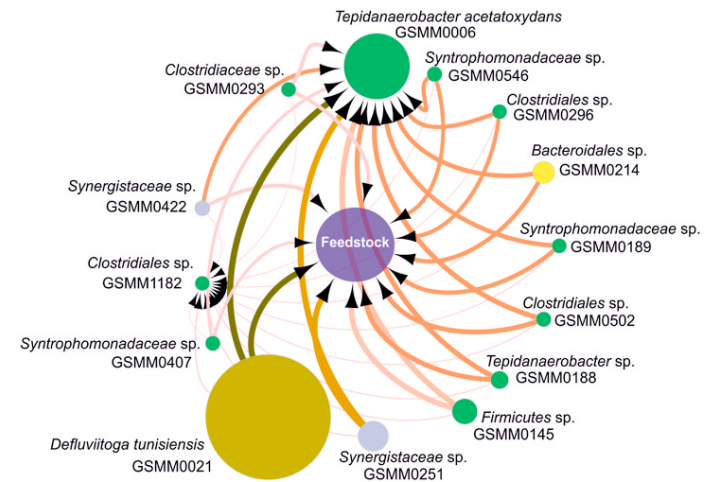
B



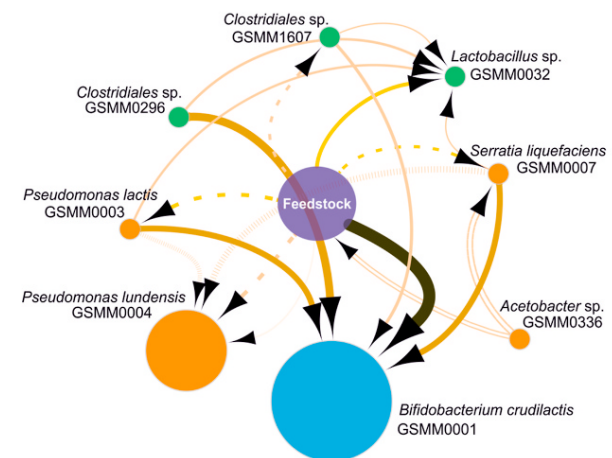


Microbial interactions in AD

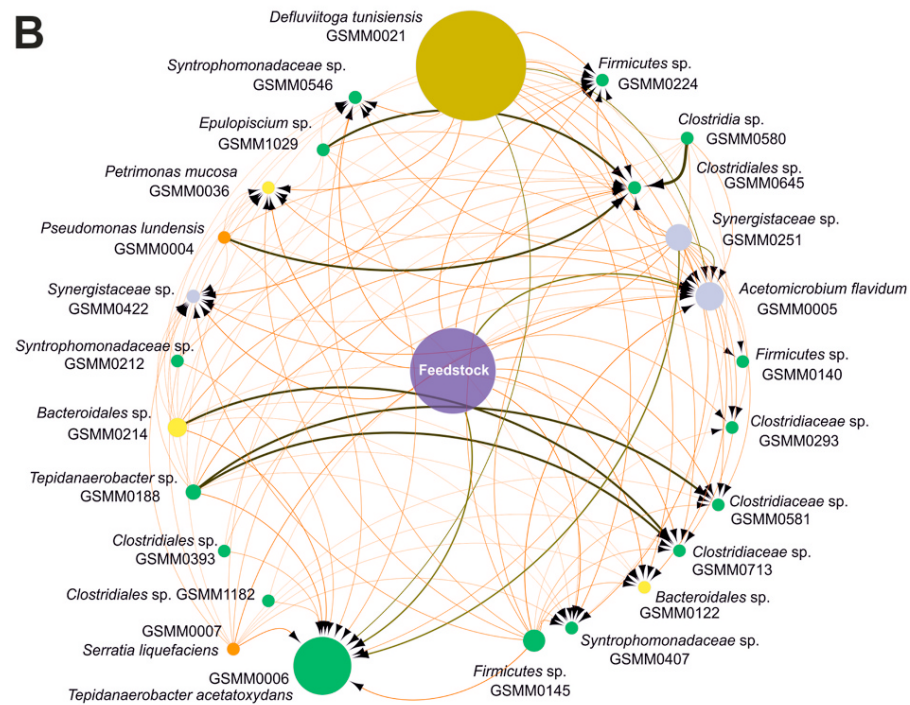
A



C



B





Final remarks



Final remarks

GSMM a holy grail?



Final remarks

GSMM a holy grail? → **not quite**



Final remarks

- Vast search space (*genes missing, or with unknown functions*)



Final remarks

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- Experimenting is computationally and time-wise intensive (*constraints, futile cycles, sinks*)



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- GSMM: what could be and NOT what is → other omics to corroborate



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- Productivity vs survival → finding the balance



Final remarks

- Vast search space (*genes missing, or with unknown functions*)
- Experimenting is computationally and time-wise intensive (*constraints, futile cycles, sinks*)
- GSMM: what could be and NOT what is → other omics to corroborate
- Productivity vs survival → finding the balance
- (Currently) unpredictable synergistic effects



Thank you for your attention!



References

Campanaro, S., Treu, L., Rodriguez-R, L. M., Kovalovszki, A., Ziels, R. M., Maus, I., Zhu, X., Kougias, P. G., Basile, A., Luo, G., Schlüter, A., Konstantinidis, K. T., Angelidaki, I. (2020). New insights from the biogas microbiome by comprehensive genome-resolved metagenomics of nearly 1600 species originating from multiple anaerobic digesters. *Biotechnology for biofuels*, 13(1), 1-18.

Basile, A., Campanaro, S., Kovalovszki, A., Zampieri, G., Rossi, A., Angelidaki, I., Valle, G., Treu, L. (2020). Revealing metabolic mechanisms of interaction in the anaerobic digestion microbiome by flux balance analysis. *Metabolic Engineering*, 62, 138-149.



GUILD-BASED METABOLIC MODELING FOR DECIPHERING ANAEROBIC BIOPROCESSES

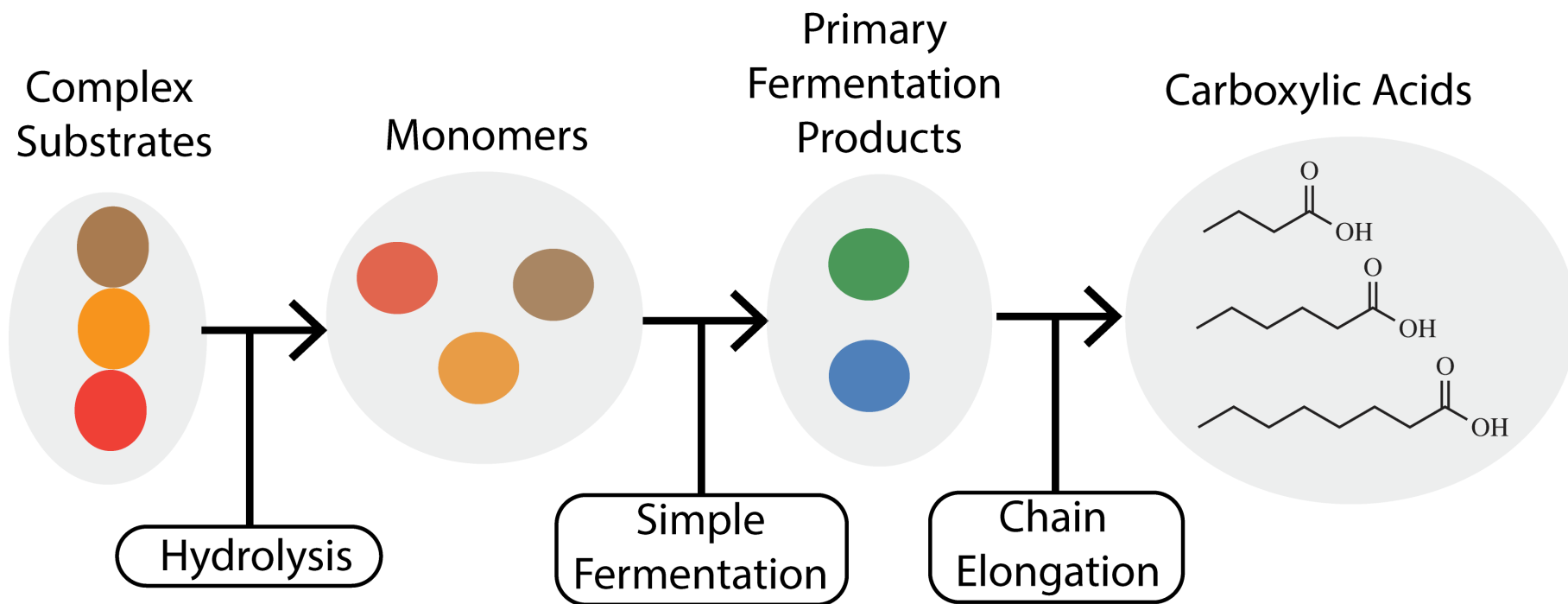
Matthew Scarborough
(Matthew.Scarborough@uvm.edu)
University of Vermont



Guild-based metabolic modeling for deciphering anaerobic bioprocesses

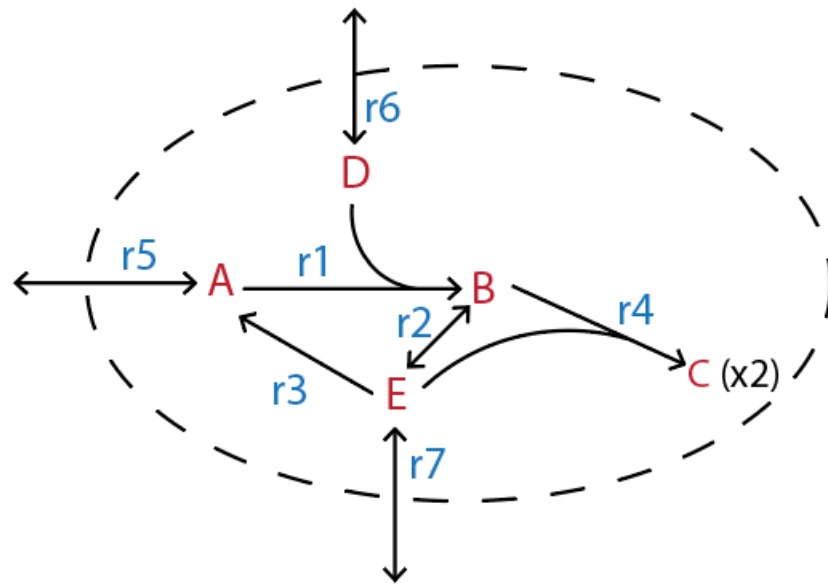
Matthew Scarborough, P.E., Ph.D.
Department of Civil and Environmental
Engineering
College of Engineering and Mathematical
Sciences
The University of Vermont

Chain elongation from complex organic wastes



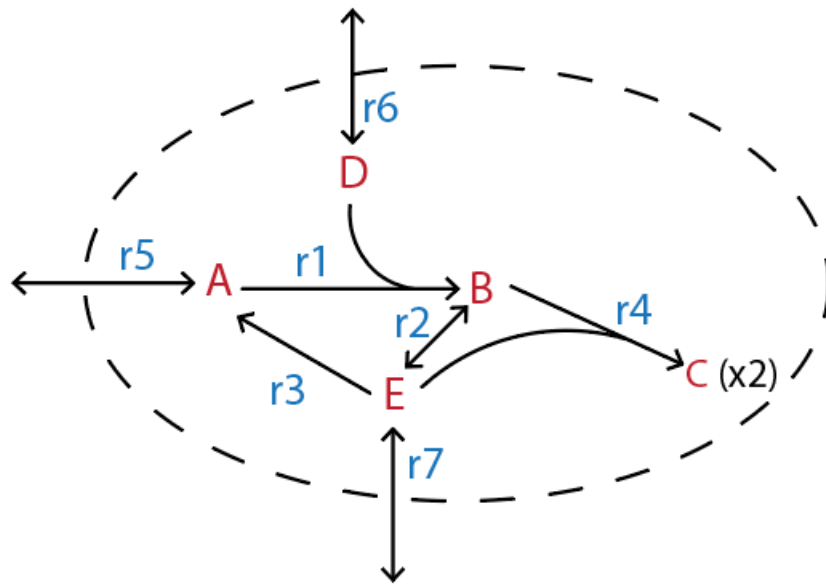
A metabolic modeling primer

Metabolic Network Representation



A metabolic modeling primer

Metabolic Network Representation



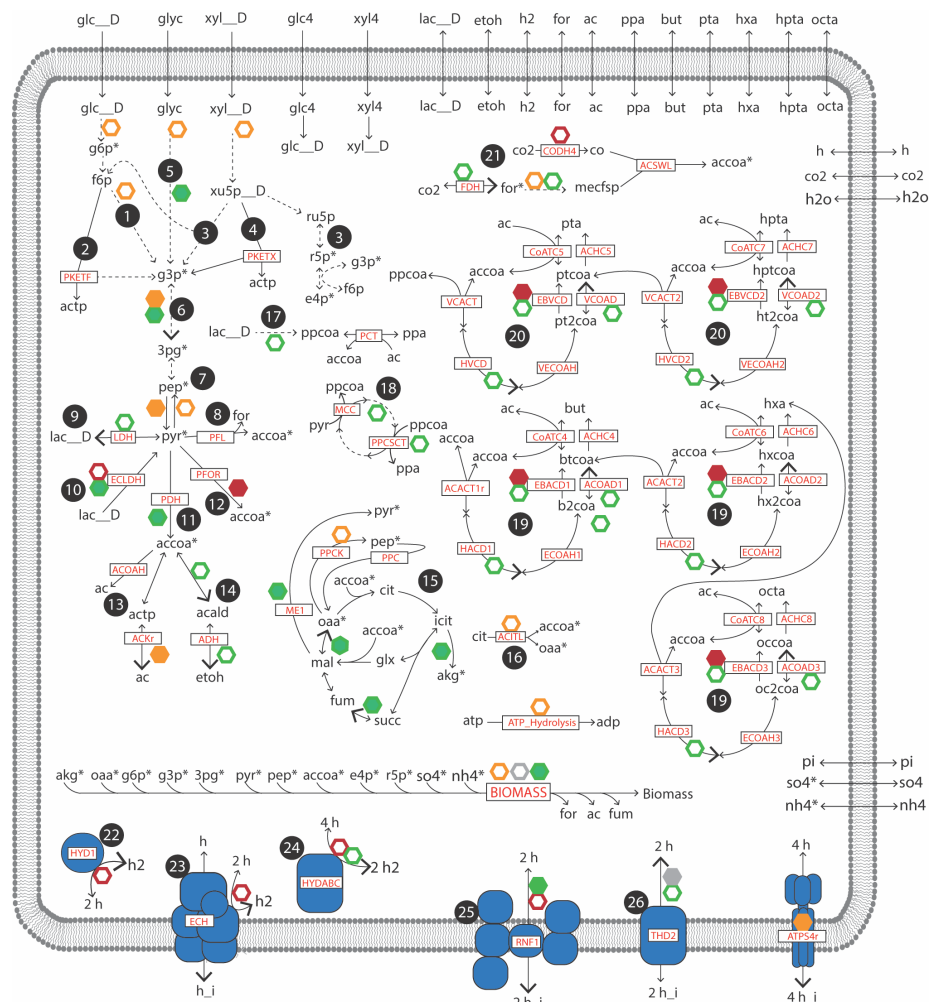
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A	-1	0	1	0	1	0	0
B	1	-2	0	-1	0	0	0
C	0	0	0	2	0	0	0
D	-1	0	0	0	0	1	0
E	0	2	-1	1	0	0	1

S

v1
v2
v3
v4
v5
v6
v7

V

$$S \bullet v = 0$$

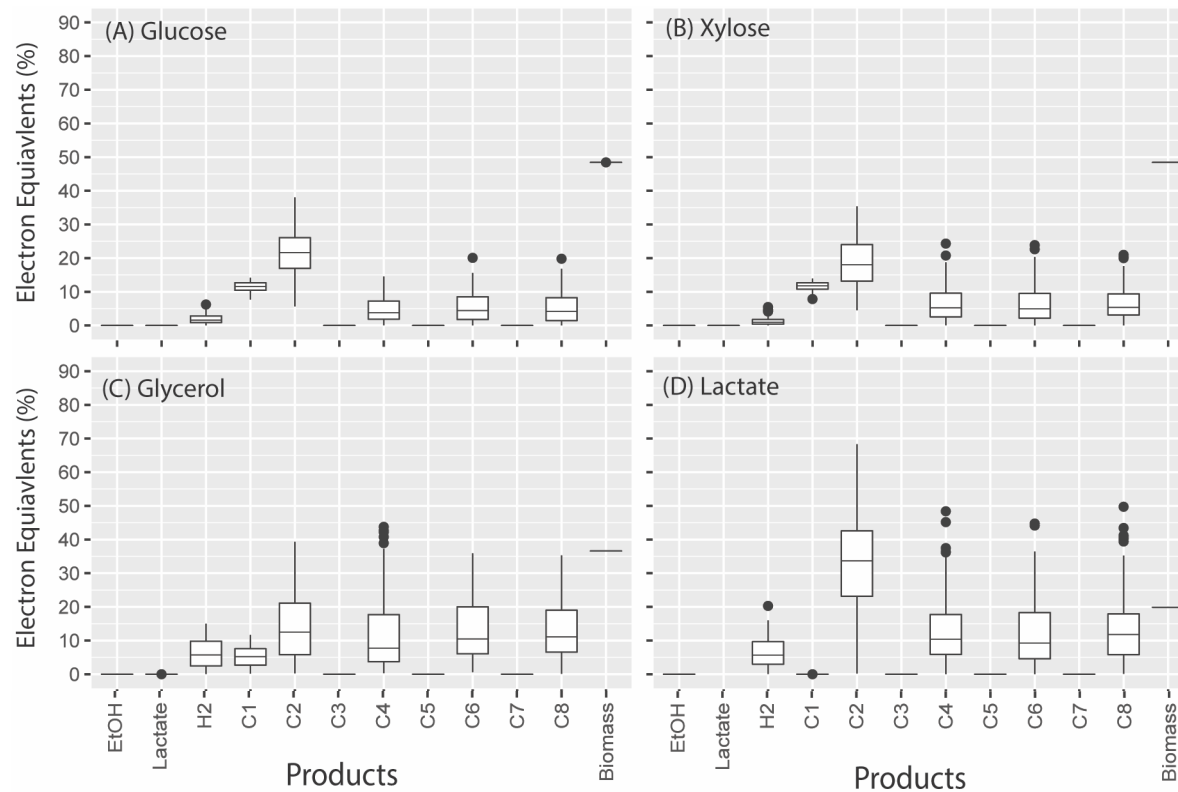


iFermCell215

Adapted from Scarborough, M, et. al. 2020. mSystems.



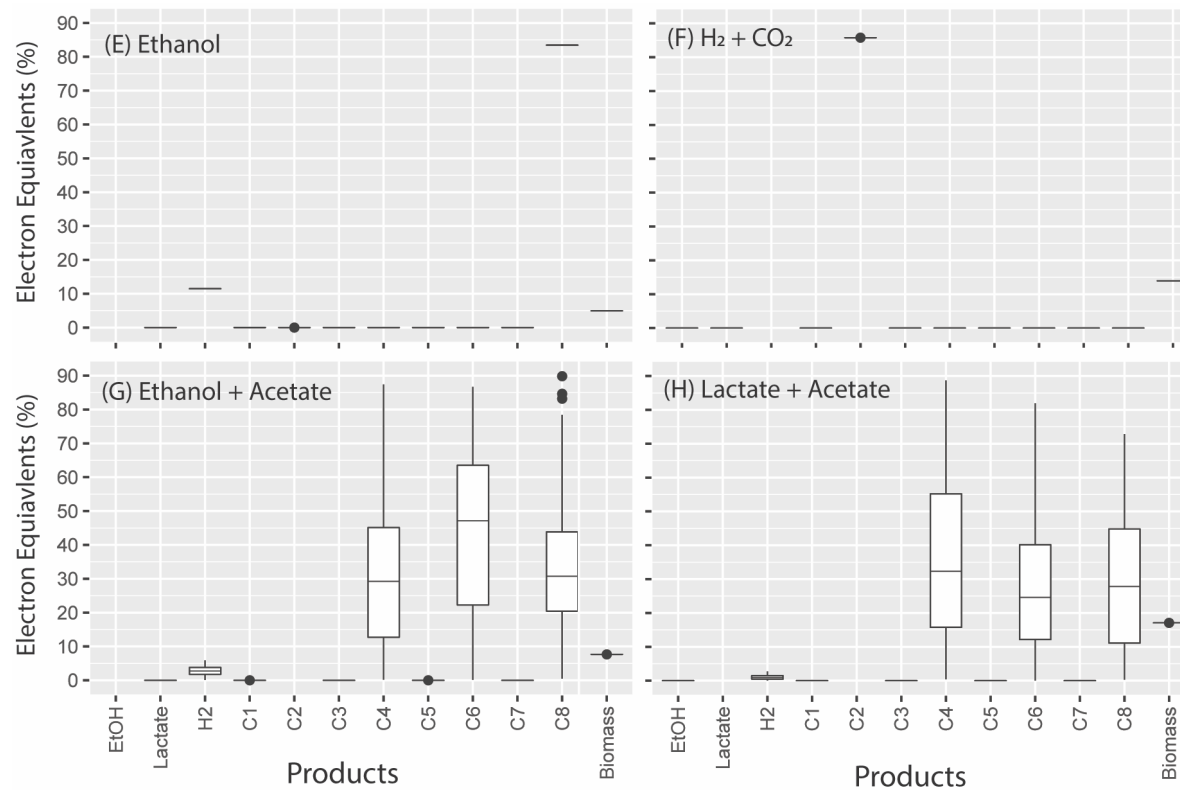
Simulating consumption of different substrates



Adapted from Scarborough, M, et. al. 2020. mSystems.



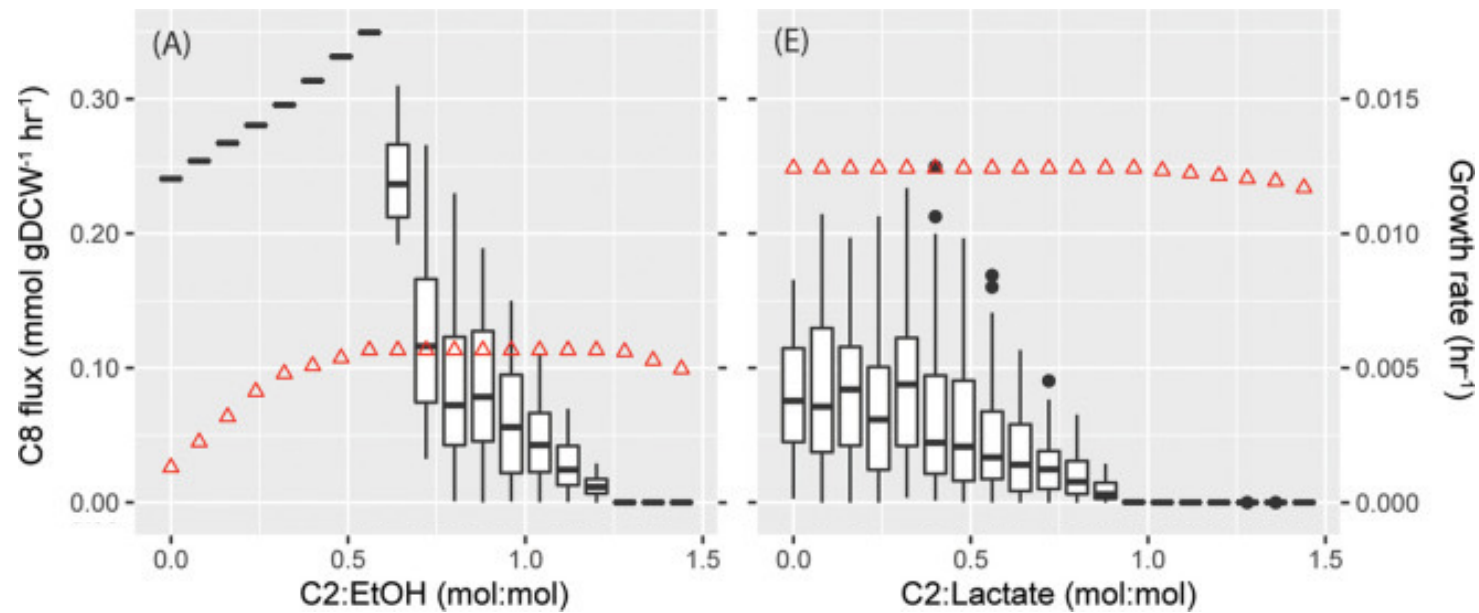
Simulating consumption of different substrates



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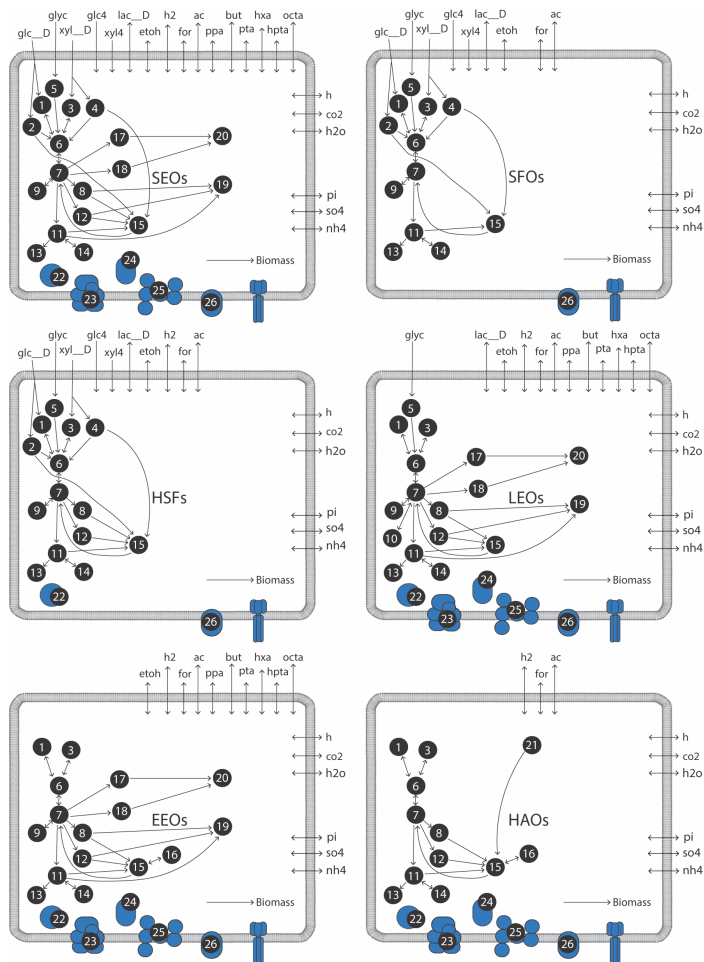


Simulating consumption of co-substrates



Adapted from Scarborough, M, et. al. 2020. mSystems.

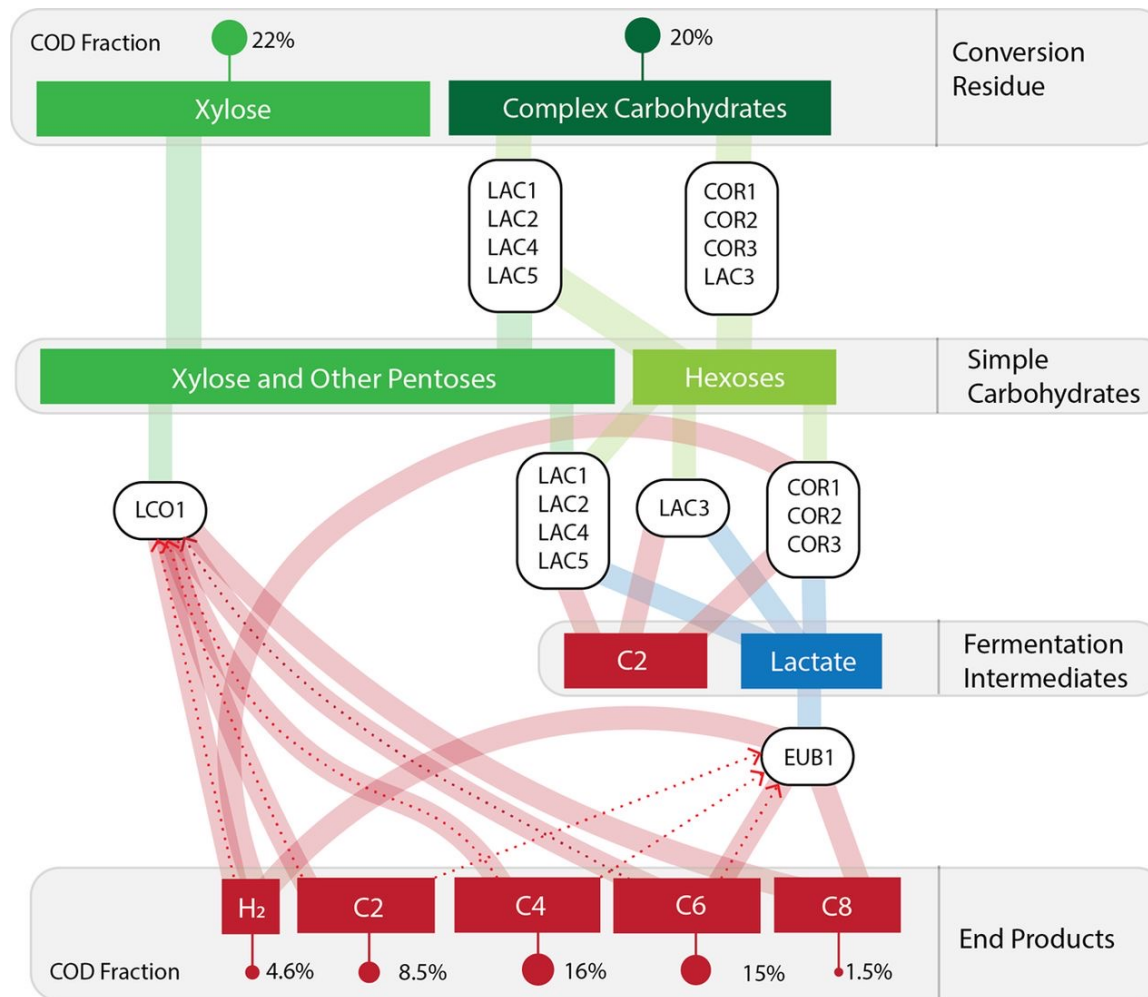


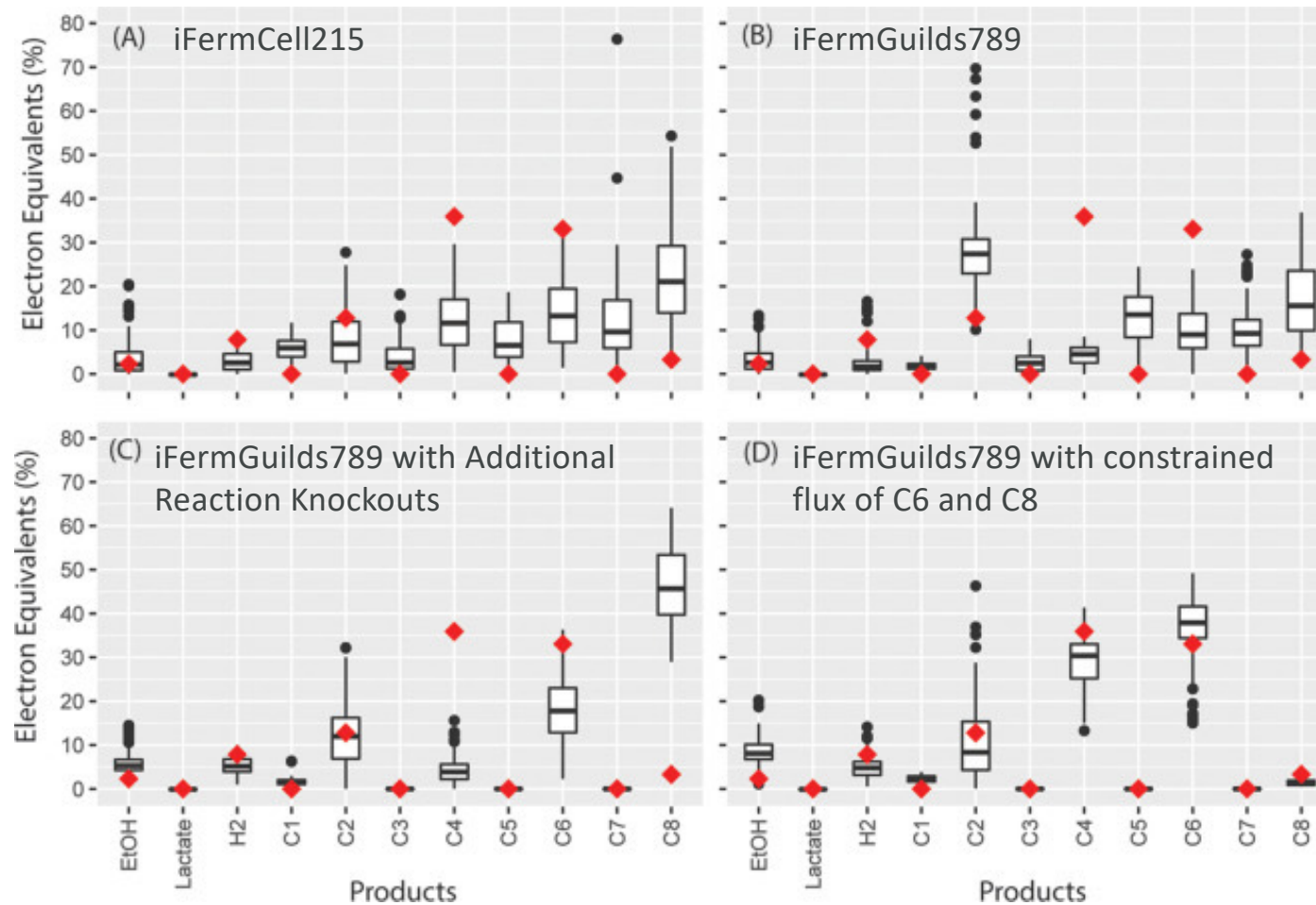


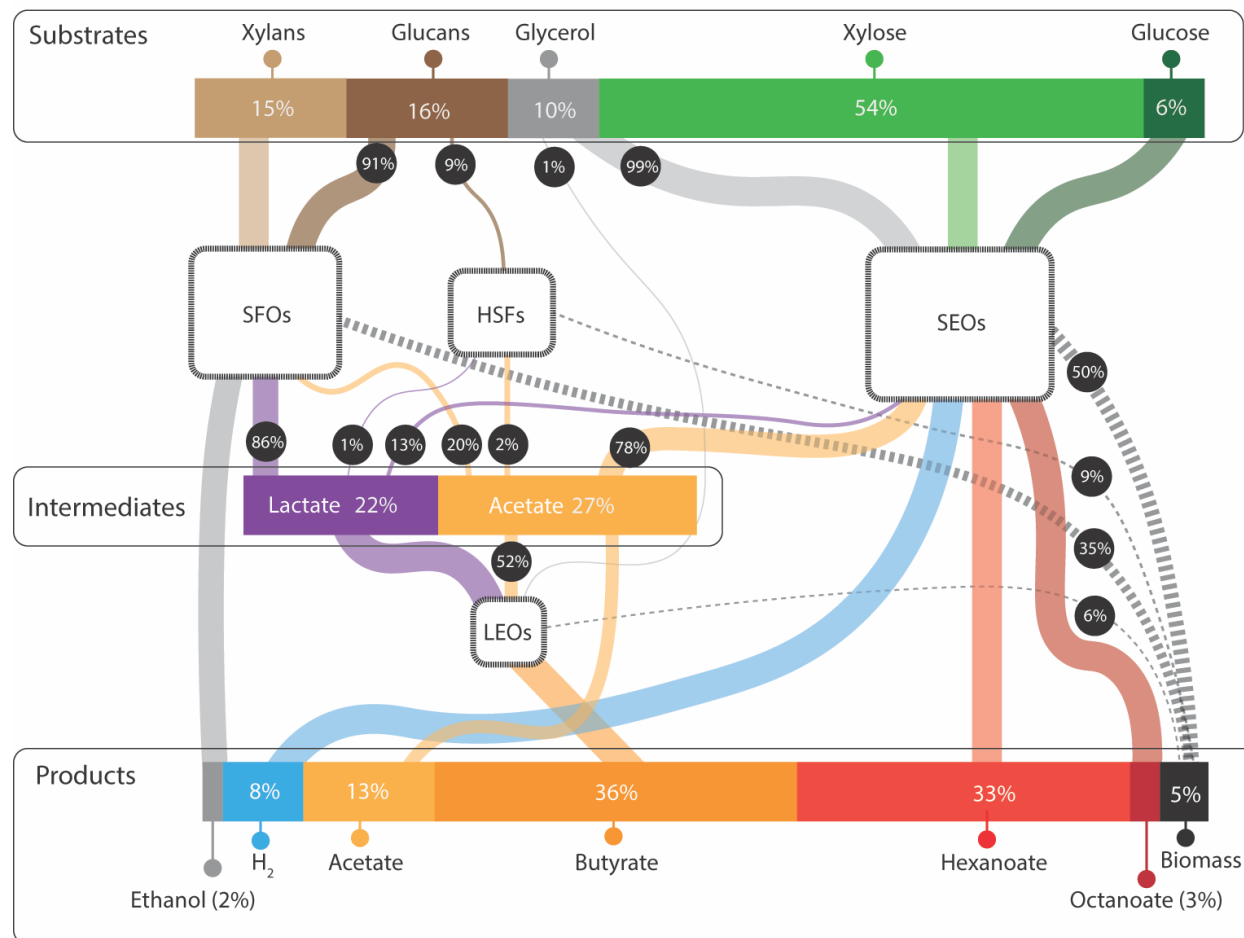
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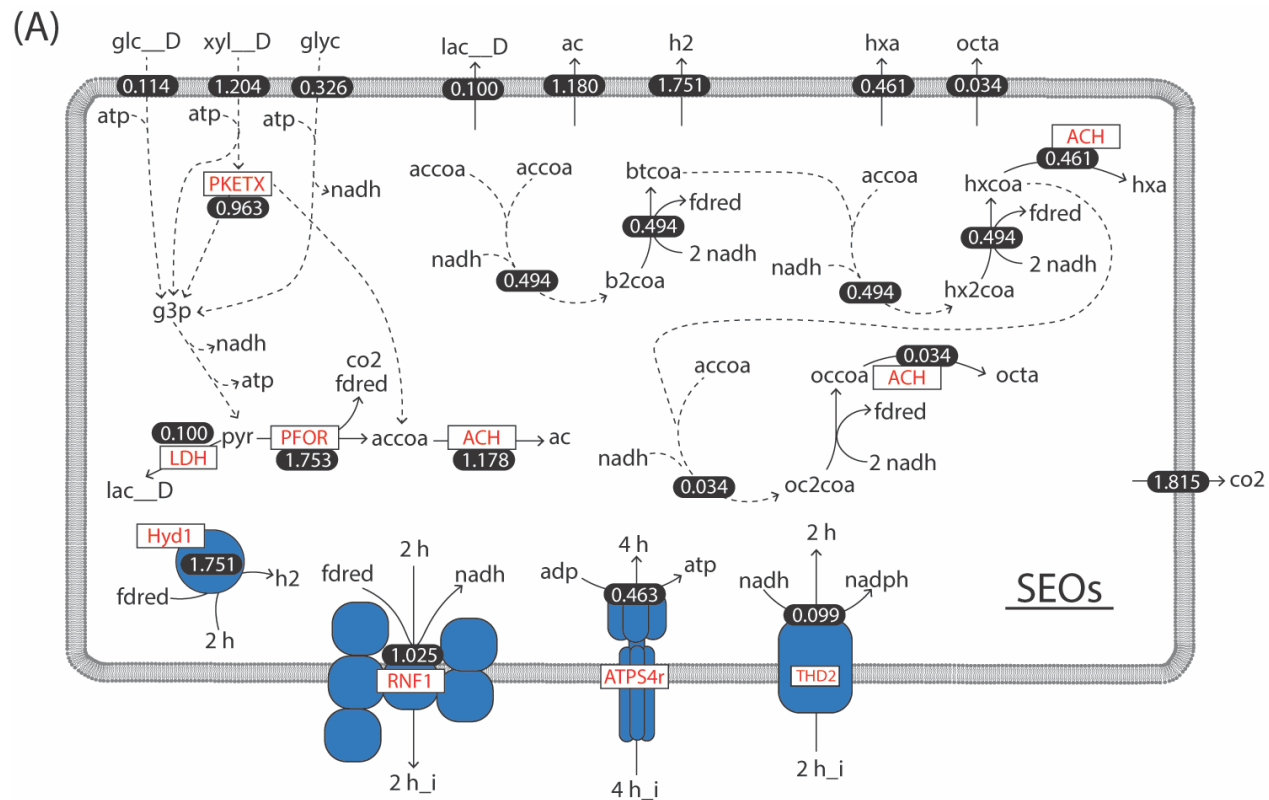


Predicting the
flow of carbon
and electrons

Adapted from Scarborough, M, et. al. 2020. mSystems.

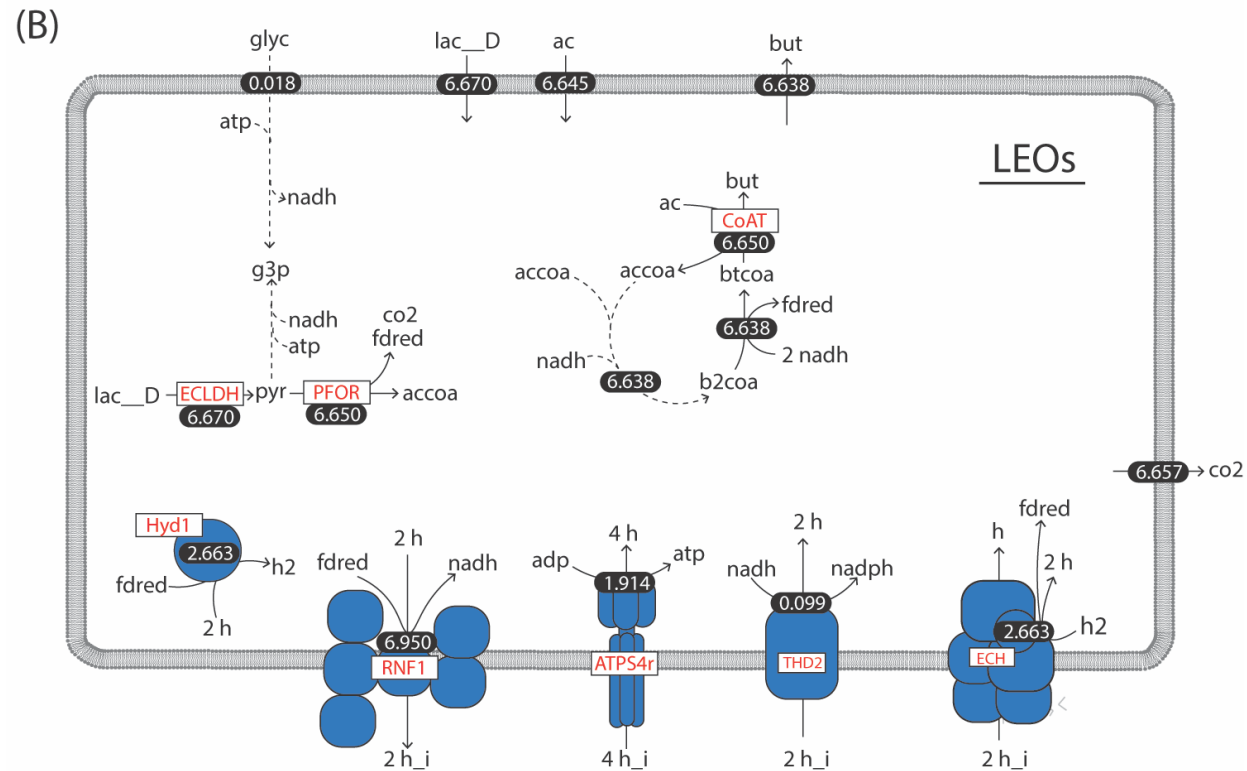


Predicted differences in sugar and lactate conversion



Adapted from Scarborough, M, et. al. 2020. mSystems.

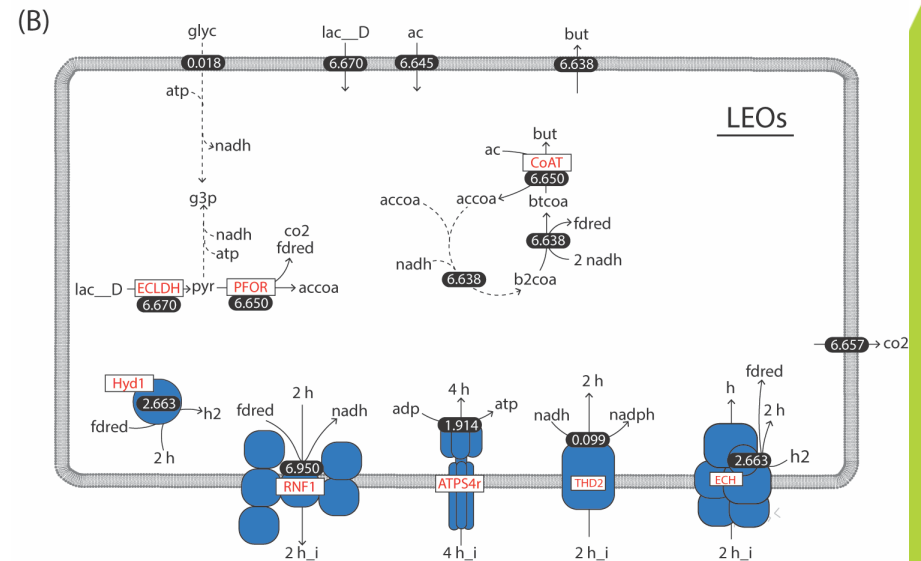
Predicted differences in sugar and lactate conversion



Adapted from Scarborough, M, et. al. 2020. mSystems.

(A)

Metabolic map of the SEOs (Secondary Electron Oxidation) pathway in *Rhodospirillum rubrum*. The map shows the conversion of various carbon sources into acetyl-CoA (ac) and then into acetyl phosphate (acp) and finally into acetyl-CoA (ac). The map includes various enzymes and their associated gene names, such as PKETX, LDH, PFOR, ACH, Hyd1, RNF1, ATPS4r, and THD2. The map also shows the conversion of various carbon sources into acetyl-CoA (ac) and then into acetyl phosphate (acp) and finally into acetyl-CoA (ac).



Adapted from Scarborough, M, et. al. 2020. mSystems.

Hypotheses from modeling

- MCFAs are produced from sugars but not lactate
- Lactate and acetate are used to produce butyrate
- Lactate consumption occurs via an electron-confurcating lactate dehydrogenase
- SEOs and LEOs vary in their routes of hydrogen production
- SEOs and LEOs vary in terminal enzymes for chain elongation

Conclusions

- Metabolic models are...
 - A useful tool to augment multi-omic techniques
 - Diagnostic tools to refine hypotheses
 - A great approach for thinking about chain elongation
- Predictive models will likely require...
 - Further understanding of chain elongator metabolism
 - Quantification of toxic effects of end-products
 - Enzyme-level kinetic analyses of different product lengths

Thanks!

mscarbor@uvm.edu

github.com/mscarbor/Mixed-Culture-Fermentation-Models



AGENDA AND HOUSEKEEPING

Speaker 1

Alberte Regueira (Universidade de Santiago de Compostela, Spain)

Speaker 2

Adam Kovalovszki (Technical University of Denmark)

Speaker 3

Matthew Scarborough (University of Vermont, USA)

Q&A Session Moderator: *Robbert Kleerebezem (Delft University of Technology, The Netherlands)*

- This session is being recorded;
- Microphones and cameras have been disabled due to the large number of attendees;
- The normal chat function is disabled;
- Please put any questions and comments you may have in the Q&A (icon to the low right in Zoom) and we will do our best to answer them during the session (in writing or orally).



CLOSING REMARKS

Great thanks to all presenters for a wonderful show!

Look out for MIA's NEXT webinar in May 2022:

“Topic to be decided”

If you have ideas for your own future webinar then contact MIA MC and we will help you make it happen!



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<http://iwa-mia.org/>

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