

A thermodynamic theory of microbial growth and its perspectives for modelling environmental biotechnology processes

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MELISSA WORKSHOP

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Microbial successions and energy gradients

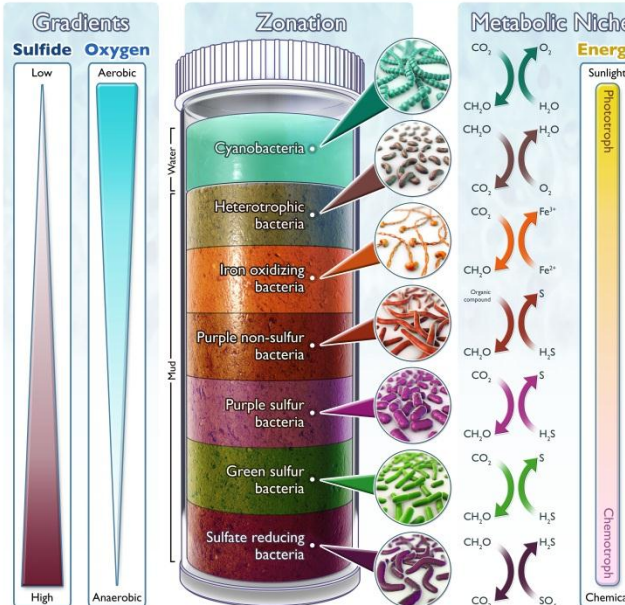
HHMI
HOWARD HUGHES MEDICAL INSTITUTE

Winogradsky Column: Microbial Ecology in a Bottle

BioInteractive

Sergei N. Winogradsky was one of the first microbiologists to study the organisms found in complex biofilm communities. One of the strategies he used to isolate organisms from nature was a miniature model of a pond cross section that is now called a Winogradsky column. It is a simple device for constructing a stratified ecosystem and provides a visual example of various modes of metabolism and zonation in the microbial world. It is a classic demonstration of the metabolic diversity of prokaryotes.

Life Environment



A soil or sediment sample is collected from nearly any source and amended with a variety of compounds such as carbon, sulfur, iron, and/or calcium. The mixture is added to a clear container and topped with water; the container is tightly capped to prevent evaporation. The column is incubated for weeks to months in well-lit conditions, thereby establishing gradients of oxygen, nutrients, and light. Different microbial taxa are adapted to different niches within these overlapping gradients, creating a stratified ecosystem defined by metabolic potential.

All life on Earth can be categorized according to an organism's carbon and energy source. Energy can be obtained from light reactions (phototroph) or chemical oxidations (chemotroph) for cellular synthesis can be obtained from carbon dioxide (autotroph) or from preformed organic compounds (heterotroph). These categories combined form the four basic life types and can be found among the bacteria within a single Winogradsky column: photoautotrophy, photoheterotrophy, chemoautotrophy, and chemoheterotrophy. Depending on our Winogradsky columns can search for many different types of bacteria. The illustration above lists some common examples.

Oxidation of organic matter

WATER	O_2 , Fe^{3+} , NO_3^- , SO_4^{2-}	Energy Yield
sediment column	$O_2 \rightarrow CO_2$	aerobic resp. 686
MUD	$NO_3^- + H^+ \rightarrow N_2$	dissimilatory nitrate red. 649
	$Fe^{3+} \rightarrow Fe^{2+}$	iron red. 300
	$SO_4^{2-} + 2H^+ \rightarrow HS^-$	sulfate red. 190
	$CO_2 \rightarrow CH_4$	methanogenesis 8.3

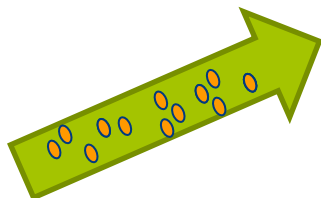
due to reduction of organic matter

This sequence also occurs in stratified lakes with anoxic hypolimnia

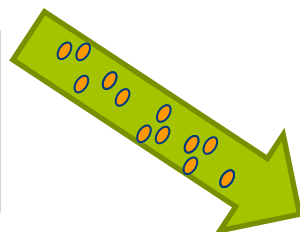
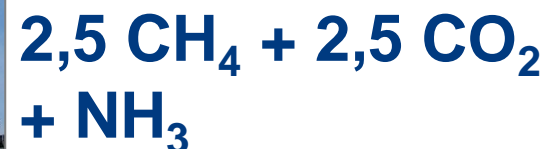
<http://www.hhmi.org/biointeractive/poster-winogradsky-column-microbial-evolution-bottle>

<http://www.esf.edu/efb/schulz/Limnology/redox.html>

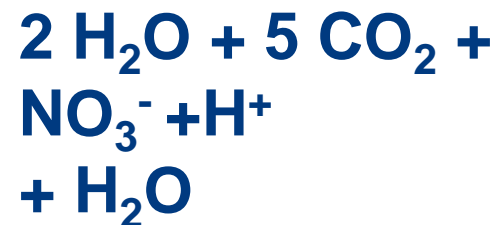
Existence of a functional convergence phenomenon of microbial communities in environmental bioprocesses



→



→



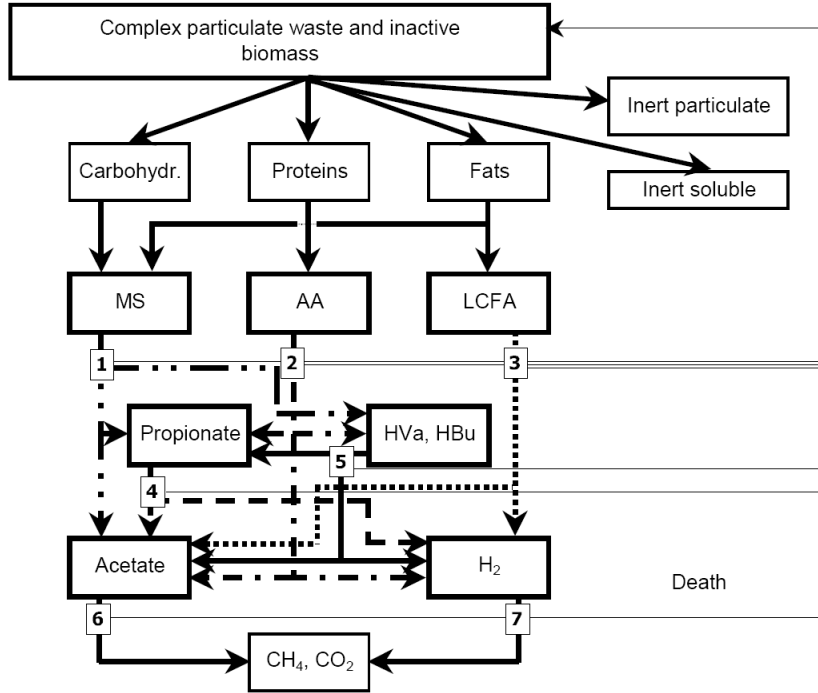
« Open diversity systems »



How could all these complex ecological interactions lead to such a reproducible functional convergence?

What do models tell us about this convergence?

Anaerobic Digestion Model N°1

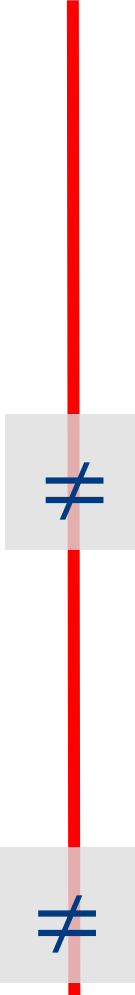


$$\frac{dB}{dt} = \mu \cdot B$$

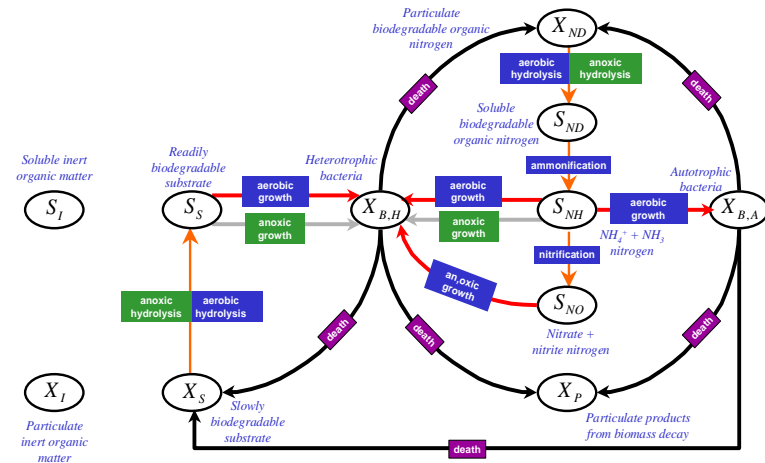
and

$$\mu = \mu_{\max} \frac{[S]}{K_s + [S]}$$

...n₁ times



Activated Sludge Model N°1



$$\frac{dB}{dt} = \mu \cdot B$$

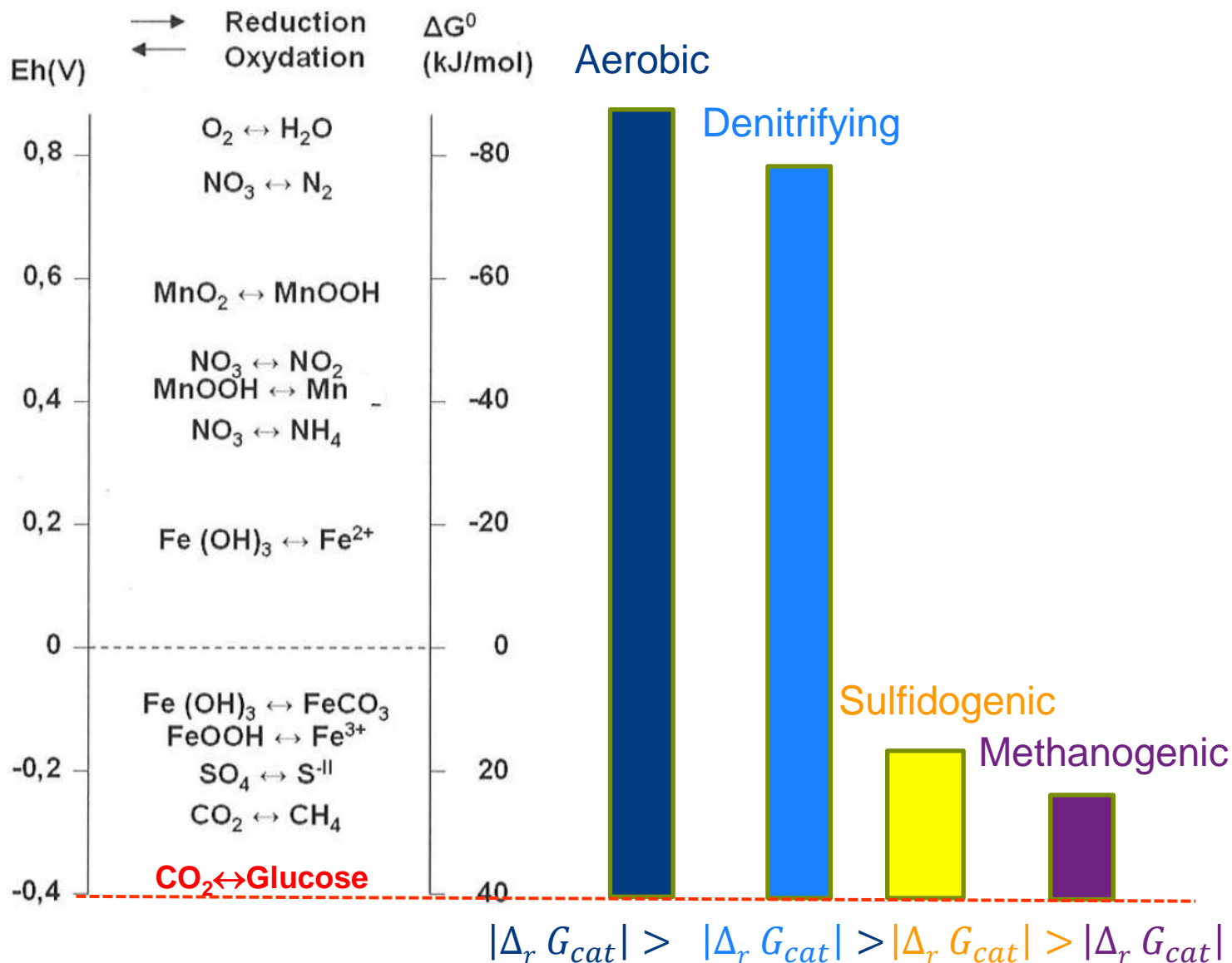
and

$$\mu = \mu_{\max} \frac{[S]}{K_s + [S]}$$

...n₂ times

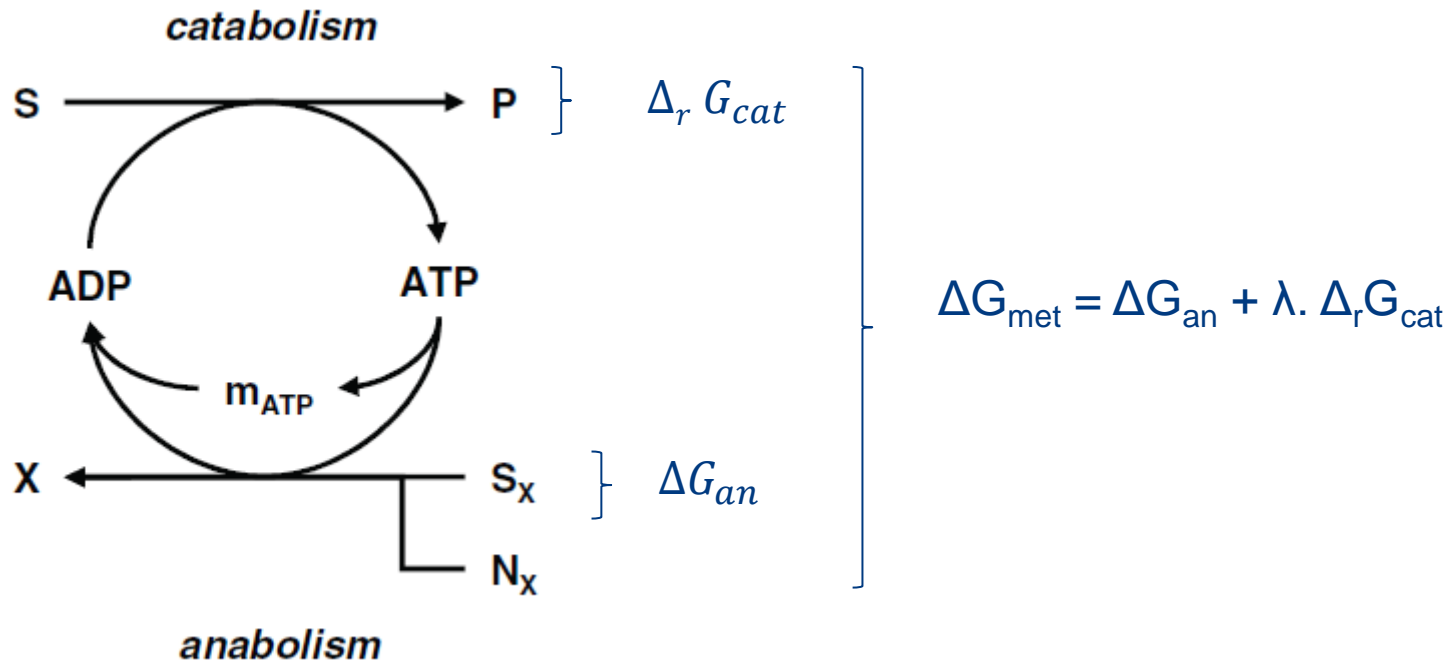
=> Current models are not appropriate to handle this question

Observed metabolic successions and thermodynamics



Thermodynamic balances of microbial growth

6



$$\Delta G_{met} = \Delta G_{an} + \lambda \cdot \Delta_r G_{cat} = \Delta G_{dis} = f(\text{substrate})$$

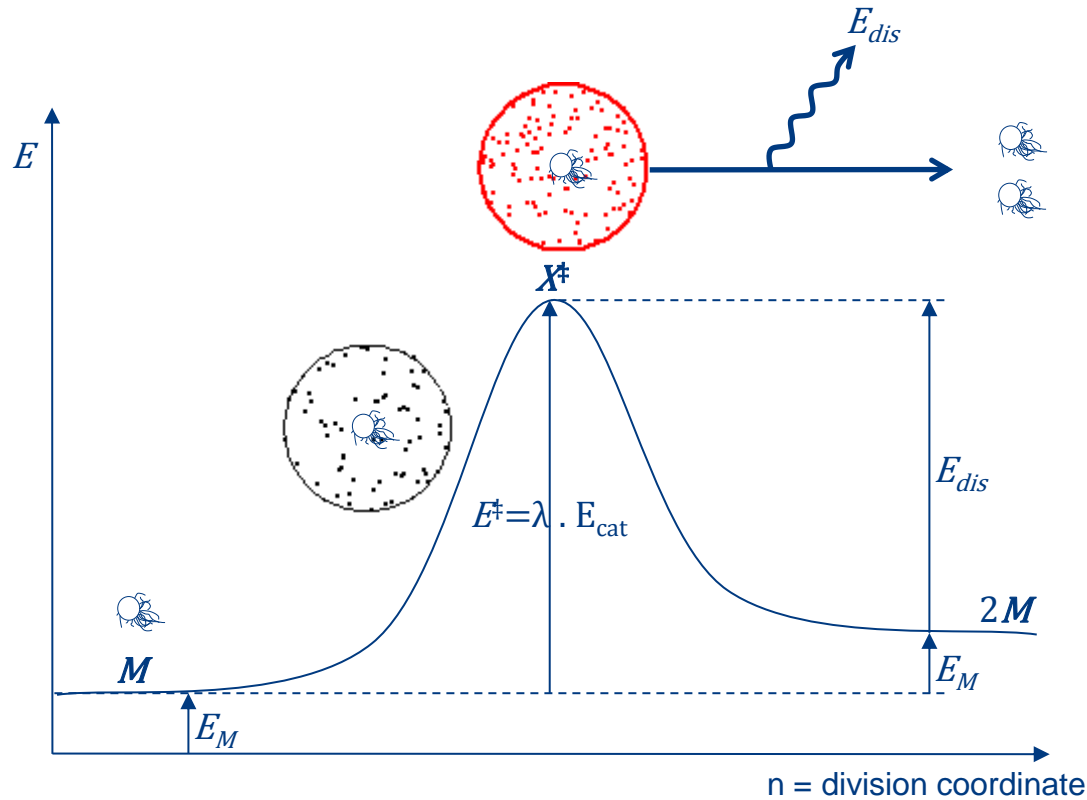
Introducing the exergy concept

$$E_{dis} = \lambda \cdot E_{cat} - E_M$$

From thermodynamic balances to kinetics?

A microbial “transition state” theory

Desmond-Le Quéméner and Bouchez, *The ISME-J*, 2014



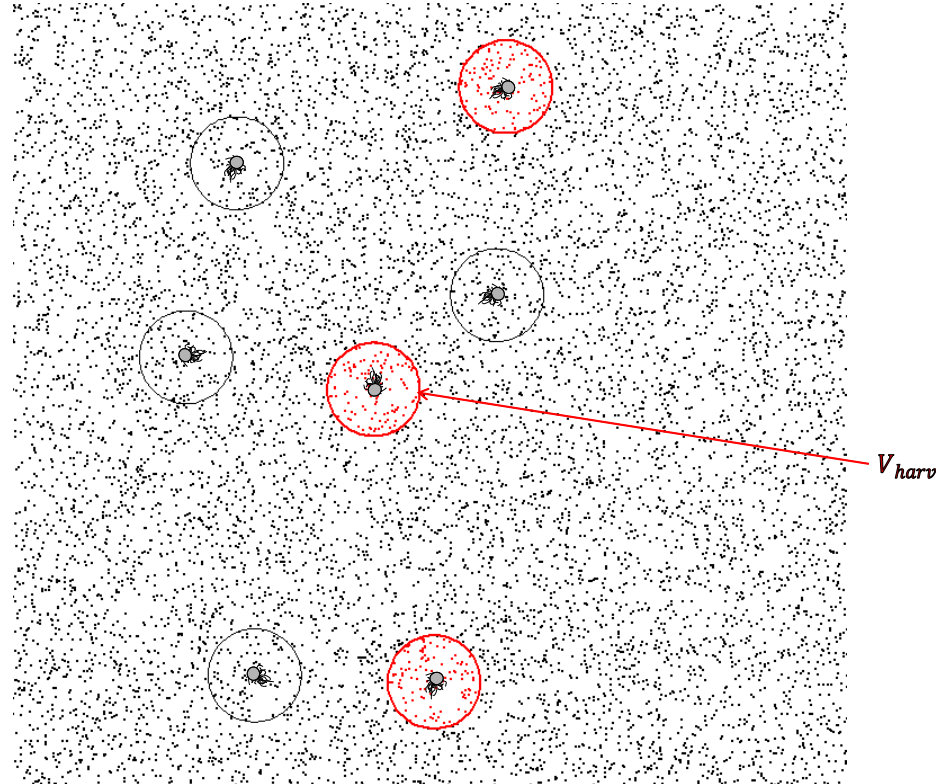
$$M \xleftarrow{K} X^\ddagger \xrightarrow{\mu_{max}} 2M$$

$$K = \frac{[X^\ddagger]}{[M]} = \frac{N^\ddagger}{N} \quad \text{and} \quad \frac{dN}{dt} = \mu_{max} \cdot N^\ddagger$$

N is the number of microbes

N^\ddagger is the number of activated microbes

Resource allocation among microbes: a statistical question



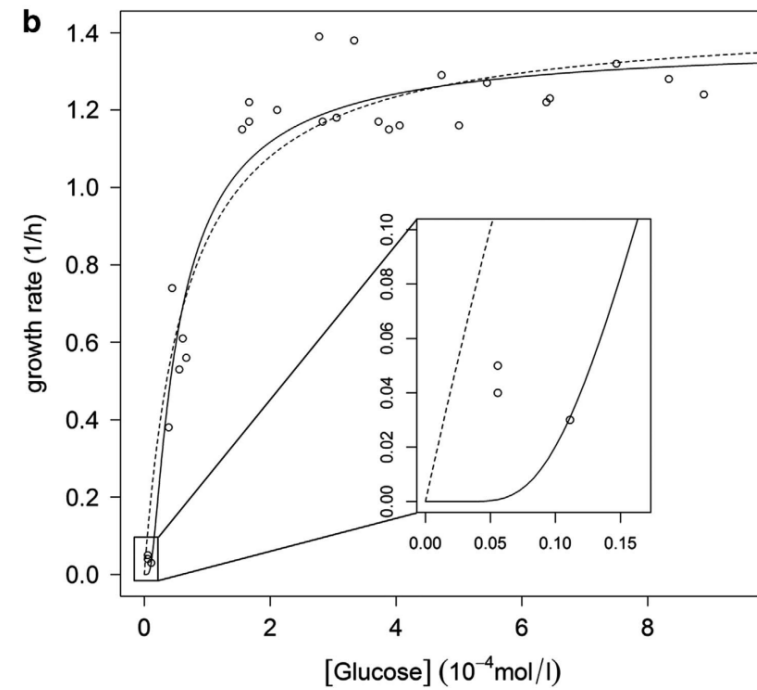
- Define the spatial distribution of molecules in the medium
 - Introduce V_{harv} « the harvesting volume »
 - Compute the distribution of molecules in the various harvesting volumes
- $\Rightarrow N^\ddagger$ can be deduced from this calculation

$$\frac{N^\ddagger}{N} = \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S] \cdot E_{cat}}\right)$$

A microbial growth equation and its consequence on isotopic fractionation

9

$$\mu = \mu_{max} \cdot \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S] \cdot E_{cat}}\right)$$



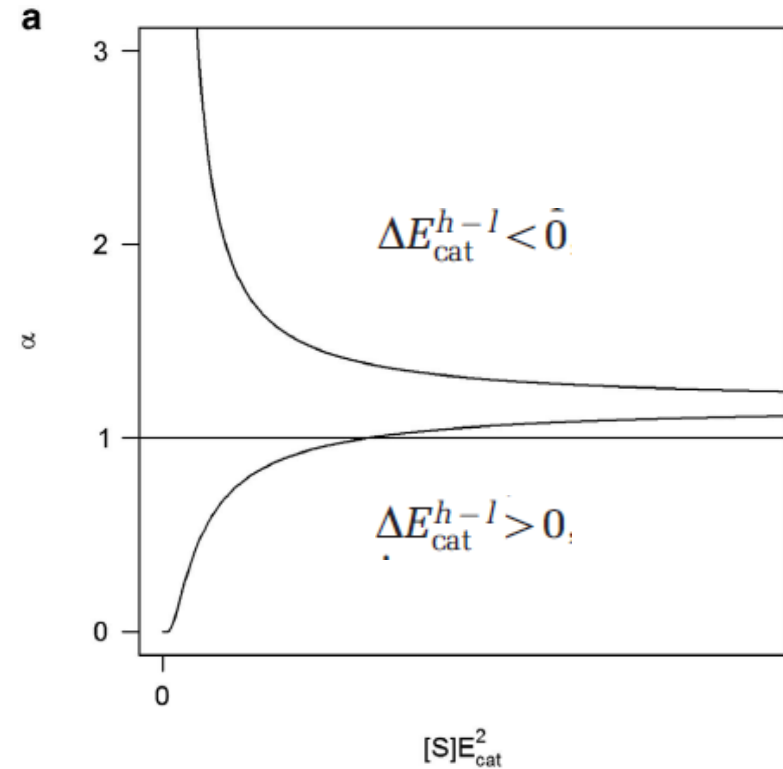
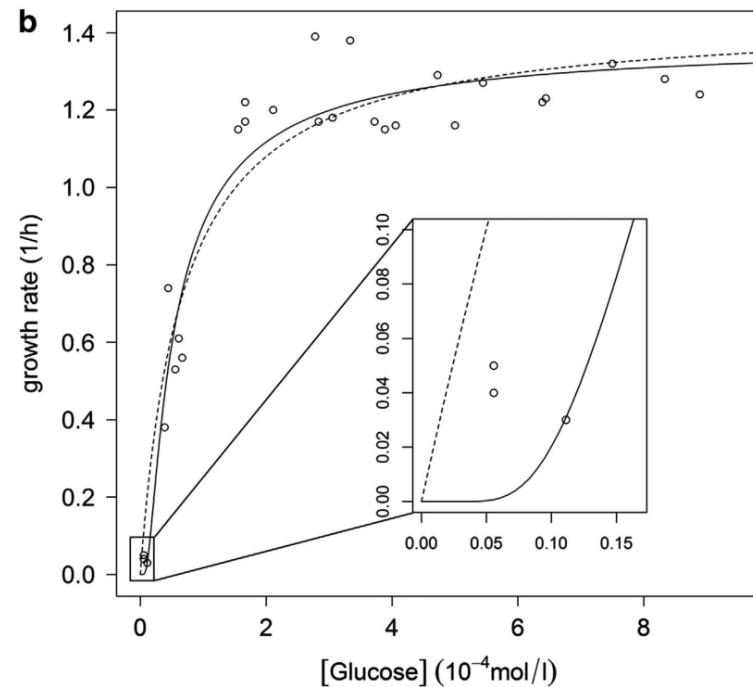
- Any further predictions?

A microbial growth equation and its consequence on isotopic fractionation

10

$$\mu = \mu_{max} \cdot \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S] \cdot E_{cat}}\right)$$

$$\alpha_{S/P} = \alpha_0 \cdot \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S]_{lim} \cdot E_{cat}^2} \Delta E_{cat}^{h-l}\right)$$

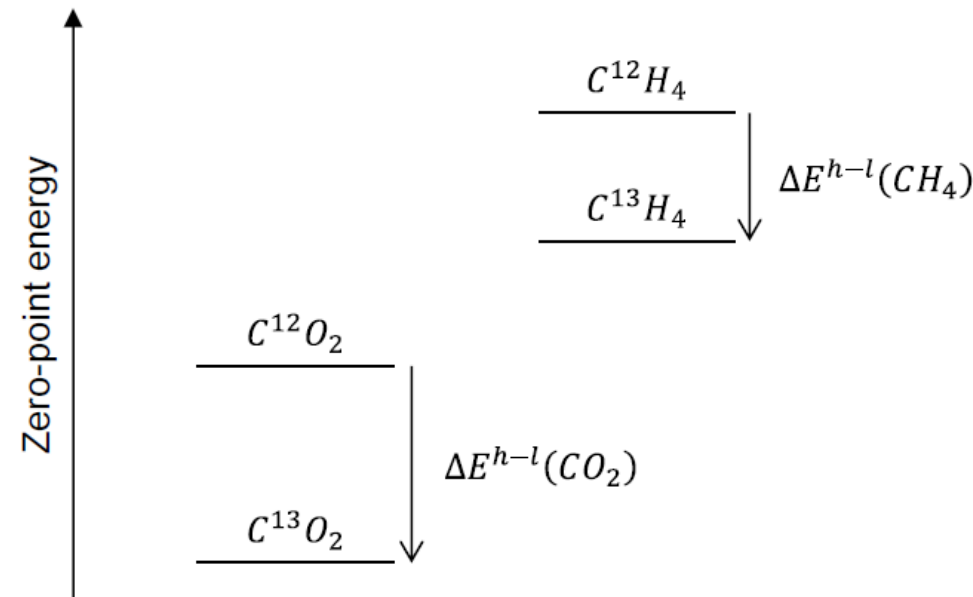
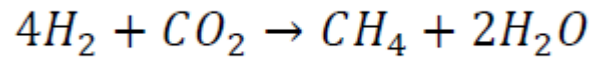


- Any further predictions?

Challenging these predictions
with real datasets...

$$\alpha_{S/P} = \alpha_0 \cdot \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S_{lim}] \cdot E_{cat}^2} \Delta E_{cat}^{h-l}\right)$$

Hydrogenotrophic methanogenesis

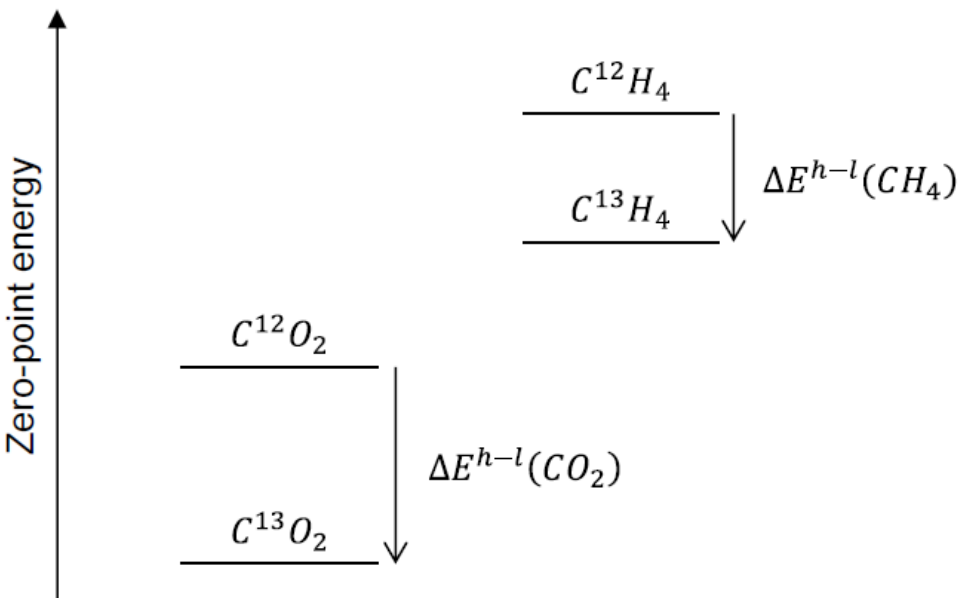
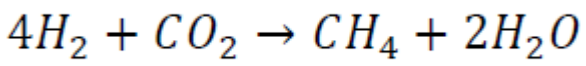


$$\Delta E_{cat}^{h-l} = \Delta E^{h-l}(CH_4) - \Delta E^{h-l}(CO_2) < 0$$

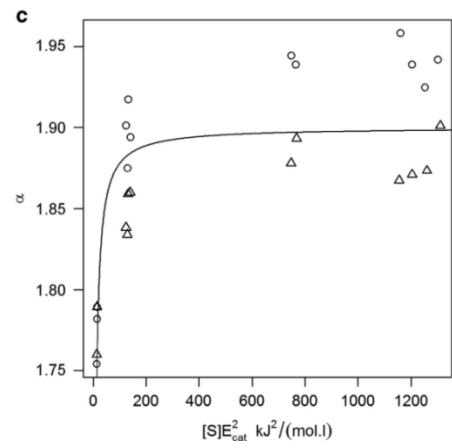
Challenging thermodynamic growth model's predictions with actual isotopic data

$$\alpha_{S/P} = \alpha_0 \cdot \exp\left(-\frac{E_M + E_{dis}}{V_{harv} \cdot [S_{lim}] \cdot E_{cat}^2} \Delta E_{cat}^{h-l}\right)$$

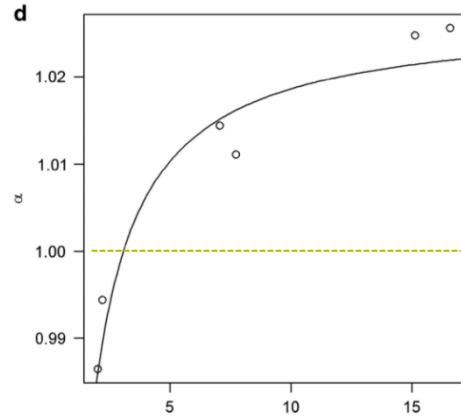
Hydrogenotrophic methanogenesis



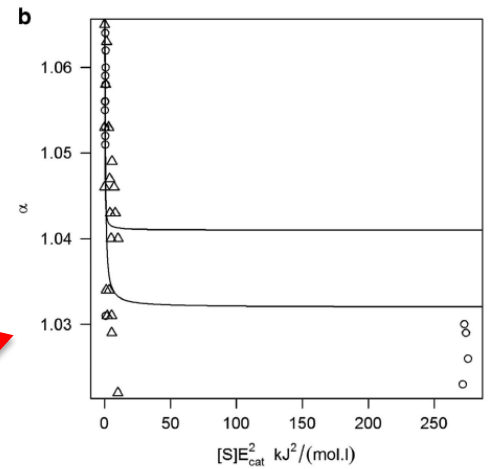
$$\Delta E_{cat}^{h-l} = \Delta E^{h-l}(CH_4) - \Delta E^{h-l}(CO_2) < 0$$



D/H
Aerobic phenol Degradation
Kampara *et al.*, 2008



$^{13}C/^{12}C$
Acetoclastic Methanogenesis
Goevert and Conrad, 2009



$^{13}C/^{12}C$
Hydrogenotrophic Methanogenesis
Valentine *et al.*, 2004
Penning *et al.*, 2005



Hadrien Delattre's PhD Project

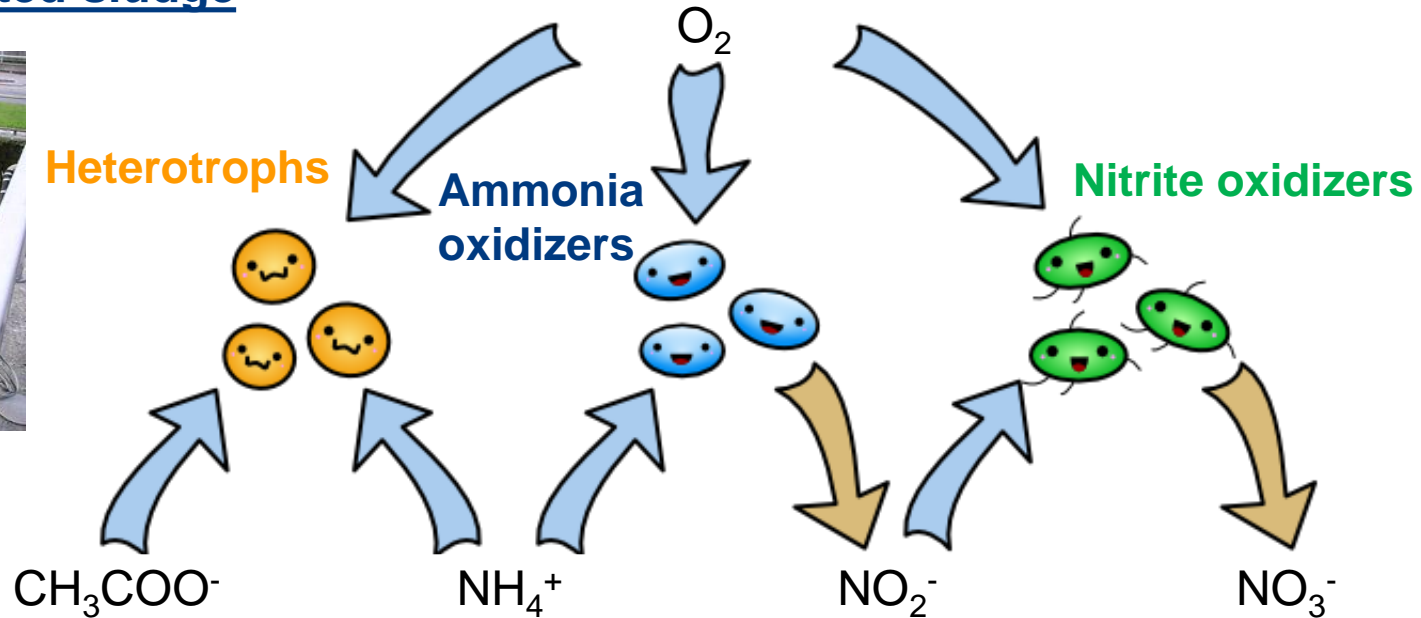
How are communities shaped by thermodynamic gradients?

$$\underbrace{\mu}_{\text{Flux: microbial growth rate}} = \mu_{max} \cdot \exp\left(-\frac{E_M + E_{dis}}{\underbrace{V_{harv} \cdot [S] \cdot E_{cat}}_{\text{Force: catabolic exergy density}}}\right)$$

Flux: microbial growth rate

Force: catabolic exergy density

Application to activated sludge

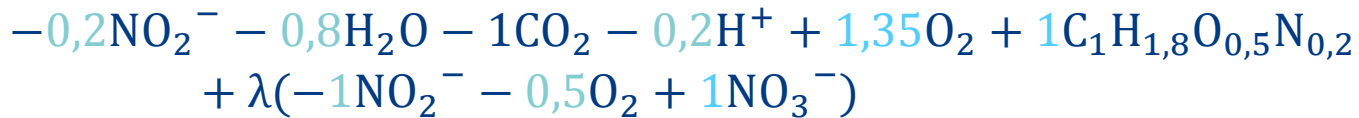
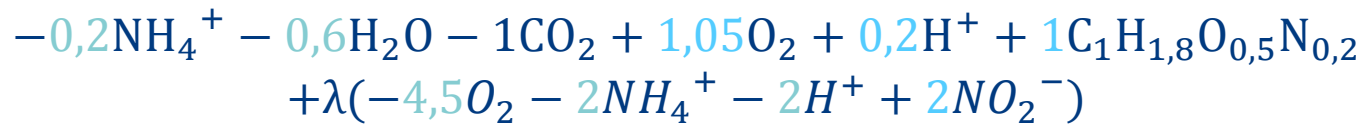
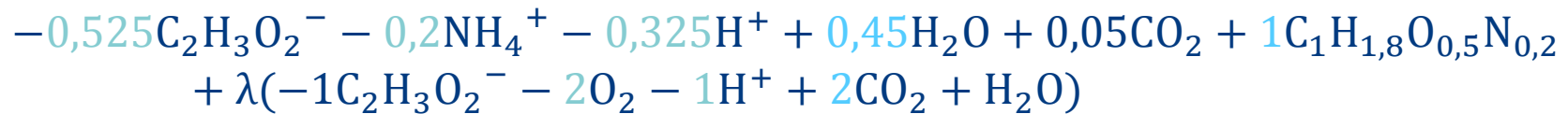
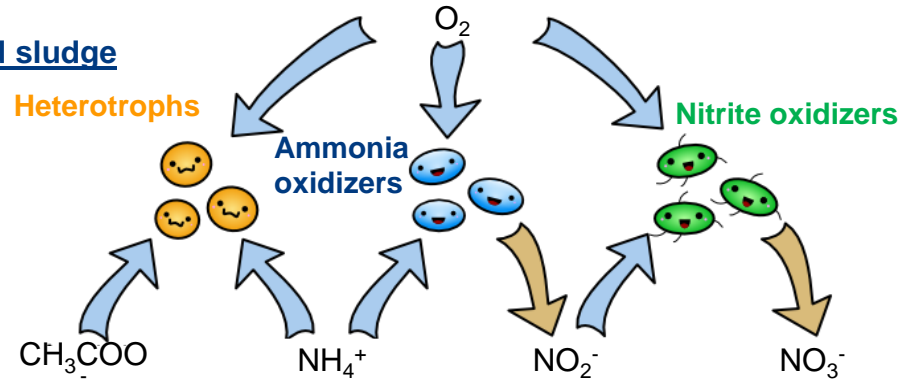


Coupled modelling of energy balances, stoichiometry and microbial dynamics



Hadrien
Delattre's
PhD Project

Application to activated sludge

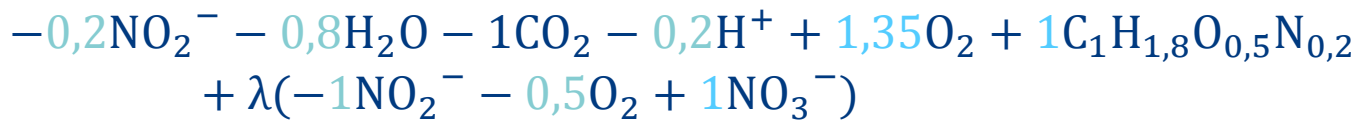
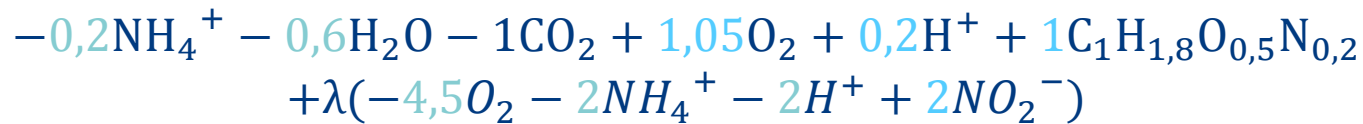
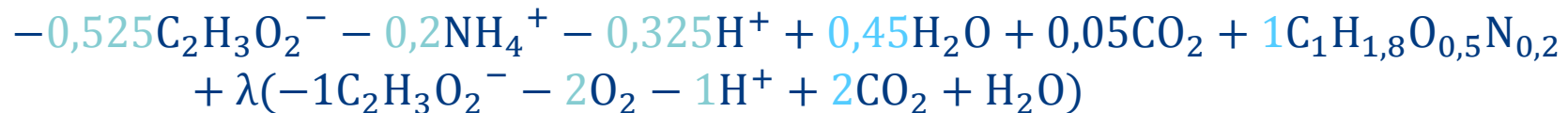
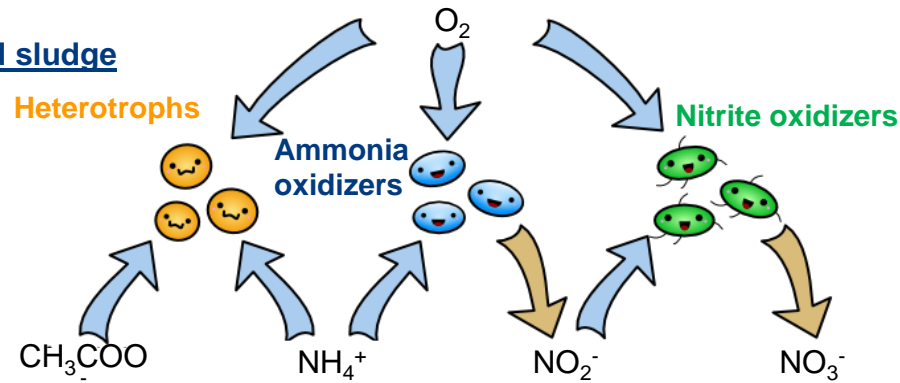


Coupled modelling of energy balances, stoichiometry and microbial dynamics



Hadrien
Delattre's
PhD Project

Application to activated sludge



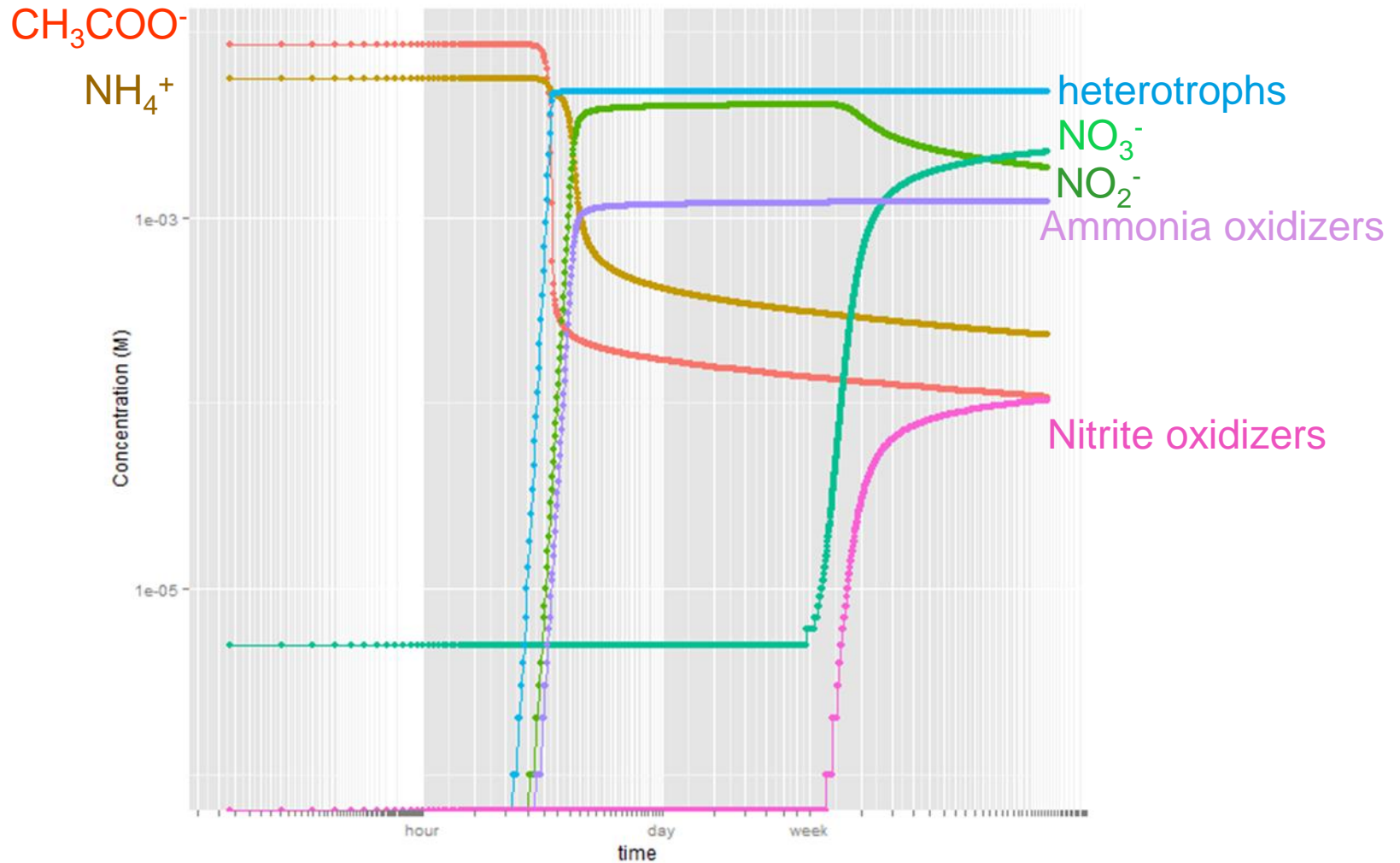
Multisubstrate rate law expression for all groups:

$$\mu = \mu_m \prod_i e^{-\frac{v_i(\lambda)}{V_h \cdot S_i}}$$

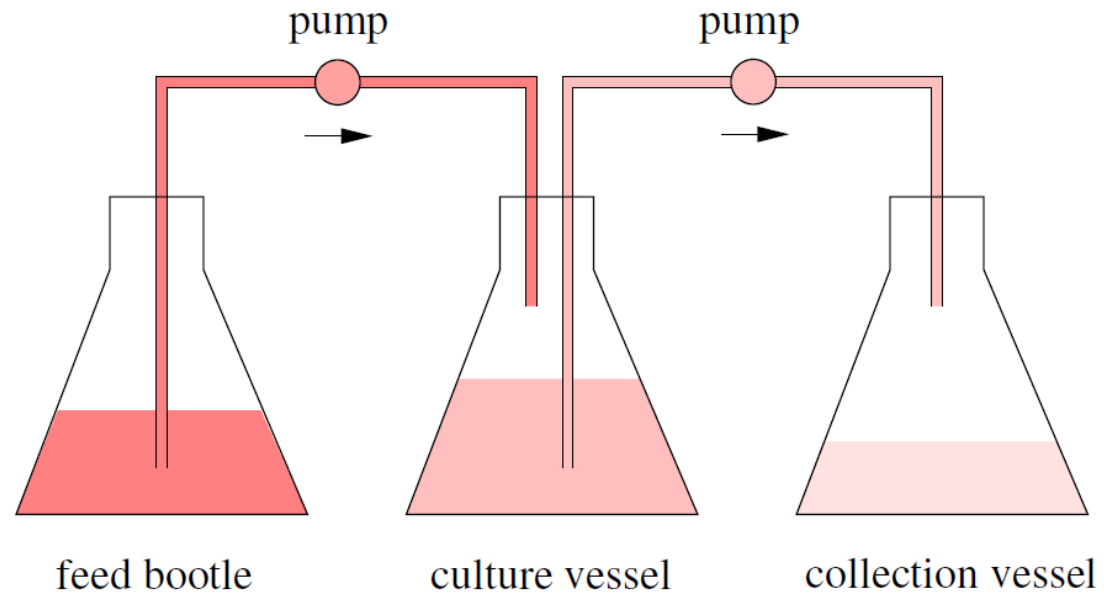
λ is calculated from thermodynamic balances of microbial growth

Simulating activated sludge batches fed with wastewater

Simulation conditions: constant supply of air, default ASM chemical species and microbial composition used for initialization, same default *E.coli* growth parameters for all groups



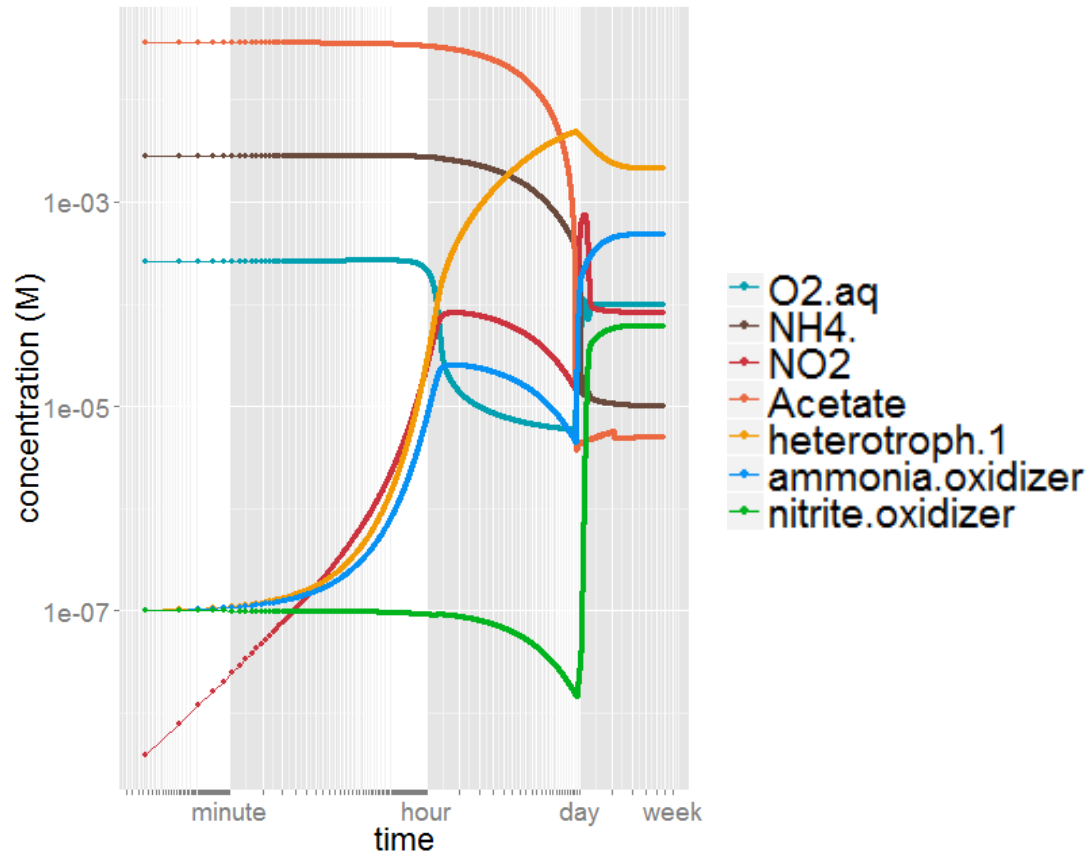
Simulating « chemostats »



Monod 1950 – Novick & Szilard 1950

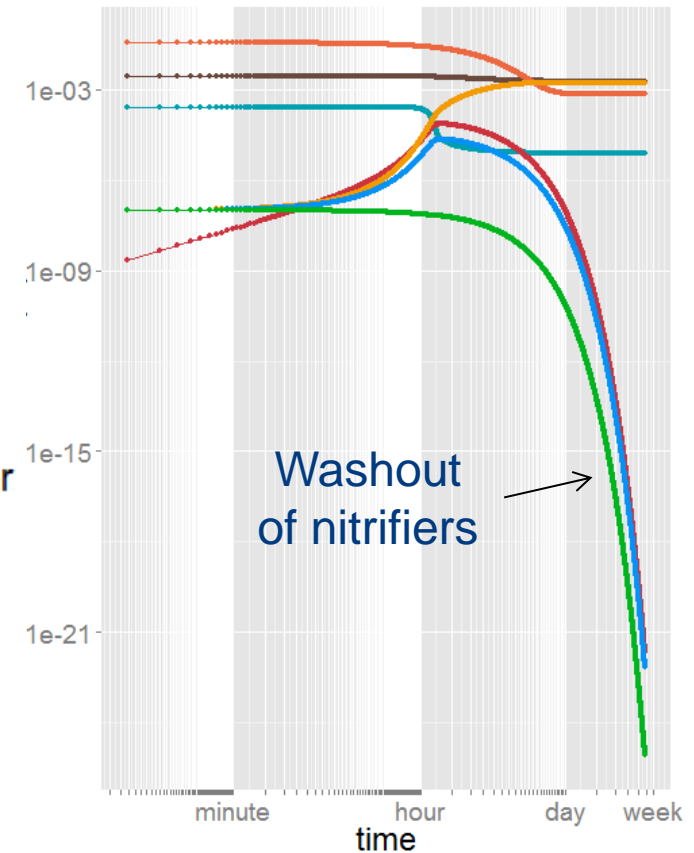
Varying the loading rate in virtual activated sludge chemostats

Low loading rate



➤ Nitrification

High loading rate



➤ No nitrification

⇒ Engineering rules as emerging properties of thermodynamic models



Take home messages...

- A consistent framework for coupled modelling of energy balance, stoichiometry and microbial dynamics
- Revisiting the importance of energy drivers for shaping microbial community structures
- Reproducing microbial successions without parameter fitting
- Towards more **predictive** models?

Many thanks to...

All the BIOMIC team members in Irstea-Antony

<http://www.irstea.fr/la-recherche/themes-de-recherche/tes/biomic>



Elie Desmond,
postdoc
Microbial
thermodynamics



Hadrien
Delattre, PhD,
Thermodynamic
simulation of
microbial
ecosystems



Christian Duquennoi,
multiphysic models



Ahlem Filali,
Activated sludge models

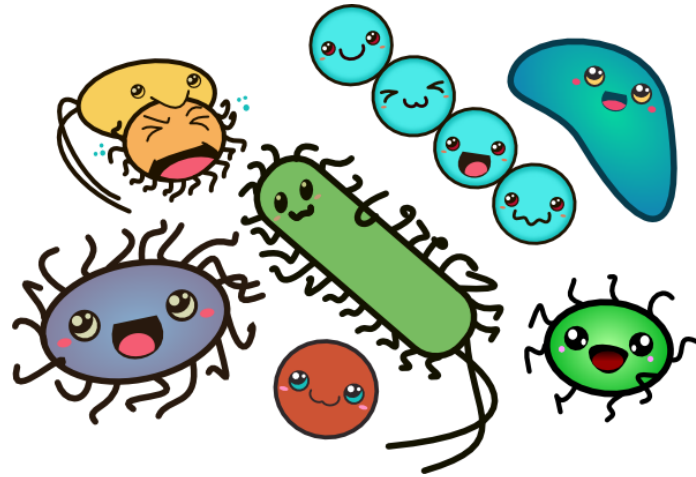
Agence Nationale de la Recherche
ANR



BIORARE project
ANR-10-BTBR-02



Thanks for you attention



Choice of a generic biomass formula

- Hoover and Porges' biomass: $C_5H_7O_2N$
 - Widely used in wastewater bioprocesses modelling
 - Gibbs energy of formation is unknown
 - Composition appears to be unrealistic
- Heijnen's biomass: $C_1H_{1.8}O_{0.5}N_{0.2}$
 - Inferred from growth data of multiple microbes on multiple substrates
 - Gibbs energy of formation is known
 - Composition appears to be unrealistic
- Battley's biomass: $C_1H_{1.613}O_{0.557}N_{0.158}$
 - Experimentally measured from *Saccharomyces cerevisiae* cultures
 - Gibbs energy of formation directly measured by calorimetry
 - Composition seems to be more realistic

- Hoover S.R. and Porges N. (1952). Assimilation of dairy wastes by activated sludge. II. The equations of synthesis and rate of oxygen utilization. *Sew. Indus. Wastes J.*, 24, 306-312.
- Heijnen J. and Dijken J. (1991). In Search of a Thermodynamic Description of Biomass Yields for the Chemotrophic Growth of Microorganisms. *biotechnology and bioengineering*, 39(8), 833-858.
- Battley E. (1998). The development of direct and indirect methods for the study of the thermodynamics of microbial growth. *Thermochimica Acta*, 309, 17-37.

Estimation of the harvesting volume (V_h)

$$\mu = \mu_m e^{\frac{v}{V_h S}}$$

$$V_h = \frac{v}{S(\ln \mu - \ln \mu_m)}$$

V_h estimated for E Coli growing on various substrates ;

- Glycerol : 0.42 m³.C-mol⁻¹ (sphere of radius 14.3 μm around each cell)
- Glucose : 0.74 m³.C-mol⁻¹ (sphere of radius 19.2 μm around each cell)
- Sorbitol : 0.79 m³.C-mol⁻¹ (sphere of radius 17.7 μm around each cell)

Lambda factor computation

\bar{S} : chemical species concentrations vector (Cx1)

$\overline{\nu}_{cat}$, $\overline{\nu}_{an}$: raw stoichiometric coefficients vectors (1xC)

$$\Delta G_{cat} = \Delta G^0_{cat} - RT \overline{\nu}_{cat} * \ln \bar{S}$$

$$\Delta G_{an} = \Delta G^0_{an} - RT \overline{\nu}_{an} * \ln \bar{S}$$

ΔG_{dis} : energie to dissipate(constant depending on the carbon source)

$$\lambda = \frac{\Delta G_{an} + \Delta G_{dis}}{-\Delta G_{cat}}$$

Dissipated energy calculation

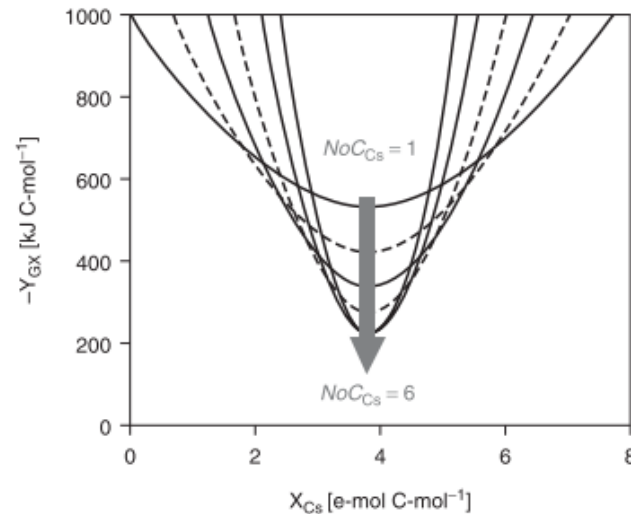


Figure 6. Minimum Gibbs energy dissipation for biomass production (Y_G^{\max} kJ/cmole X) as a function of the degree of reduction (NoC_{Cs}) and the carbon chain length of the carbon source (γ_{Cs}) according to Equation 19.

Dissipated energy is a constant whose value depends on the anabolic carbon source, according to Heijnen and Kleerebezem