

Elementary vectors for microbial communities

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40th TBI winter seminar
Bled, February 13, 2025

Joint work

Jürgen Zanghellini

Biochemical Network Analysis

Department of Analytical Chemistry

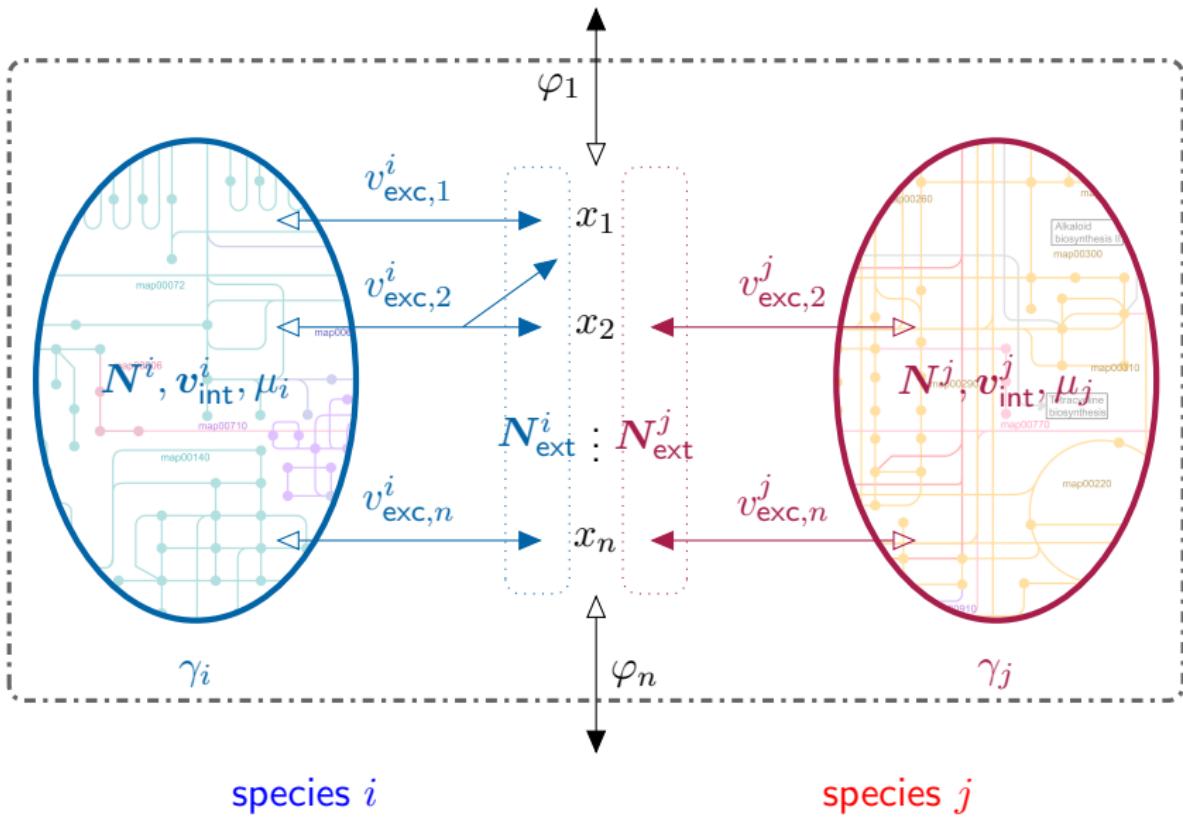
University of Vienna

S. Müller, M. Predl, J. Zanghellini

Elementary compositions of microbial communities

on arXiv/bioRxiv by end of February ...

Microbial community



Q&A

Question

Given

- a set of microbial species (their metabolic models),
- a medium, and
- a growth rate:

*What are the “minimal” communities?
And their interactions?*

Q&A

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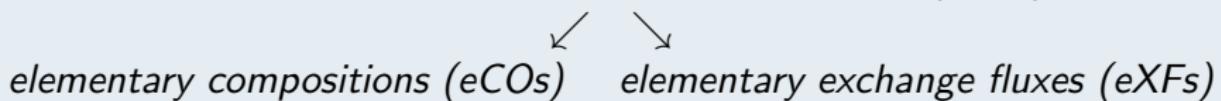
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Answer

We define:

elementary compositions & exchange fluxes (eCXs)



Metabolic networks

Stoichiometry:



metabolites A, B, C and enzyme \mathcal{E}

Kinetics:

$$v_{\mathcal{E}} = c_{\mathcal{E}} \cdot \kappa_{\mathcal{E}}(x_A, x_B, x_C; p)$$

(total) enzyme conc. $c_{\mathcal{E}} \geq 0$

metabolite conc. $x_A(t), x_B(t), x_C(t) \geq 0$

external metabolite conc., rate constants, etc. $p \geq 0$

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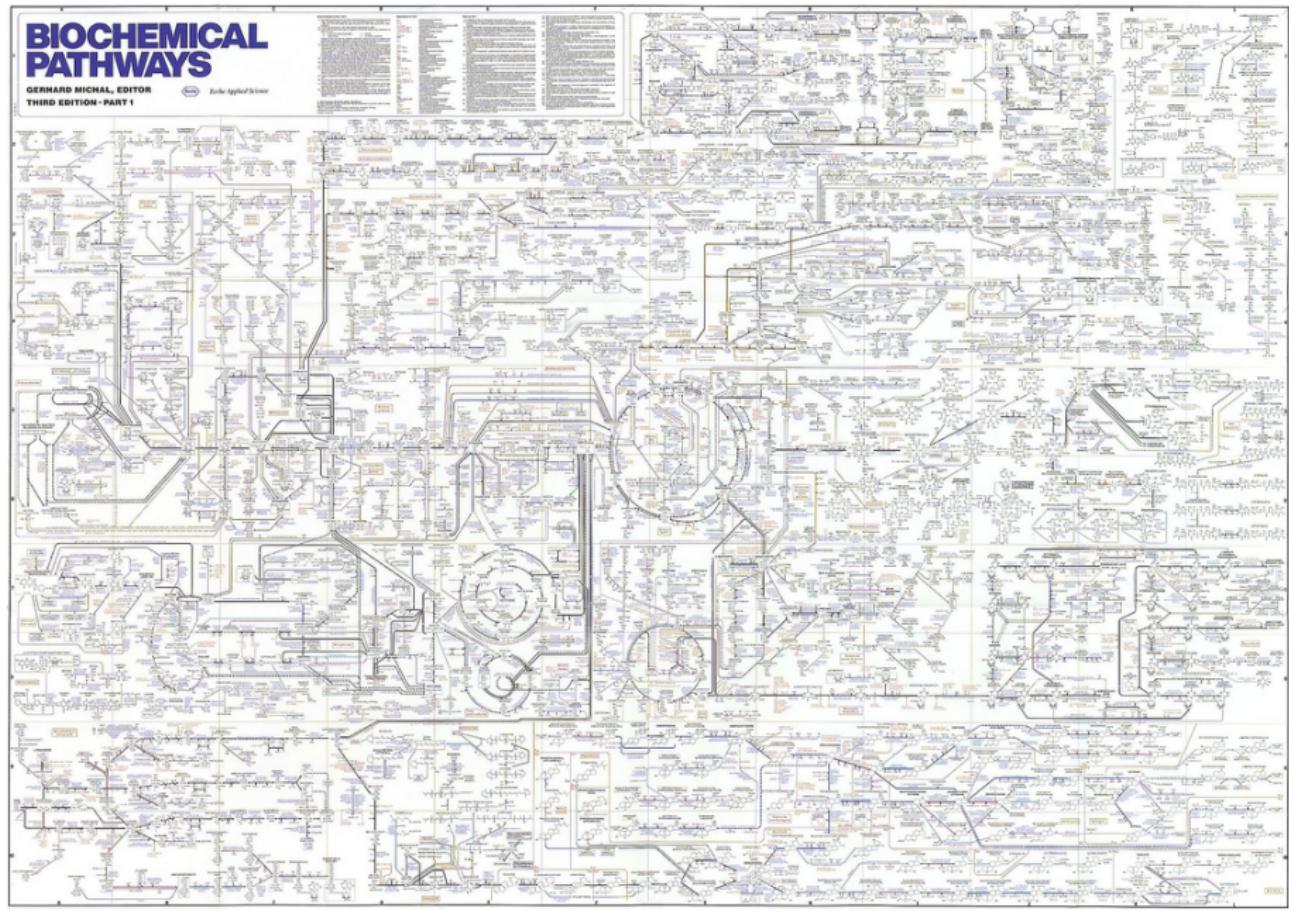
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Network dynamics:

$$\frac{d}{dt} \begin{pmatrix} x_A \\ x_B \\ x_C \\ x_D \\ \vdots \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ 1 \\ 0 \\ \vdots \end{pmatrix} v_{\mathcal{E}} + \dots, \quad \frac{dx}{dt} = Nv = N(c \circ \kappa(x, p))$$

stoichiometric matrix N , rate vector v
componentwise product \circ

Genome-scale metabolic models



Stoichiometric models

Based on $N \in \mathbb{R}^{1500 \times 2000}$, no kinetics

Flux vector v (vector of steady-state net reaction rates):

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Linear methods:

- flux balance analysis (FBA):

flux polyhedron

$$P = \{v \mid Nv = 0, v^{\text{lb}} \leq v \leq v^{\text{ub}}\}$$

$$\max_{v \in P} a^T v$$

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- elementary flux mode (EFM) analysis:

flux cone

$$C = \{v \mid Nv = 0, v_i \geq 0 \text{ for } i \in \mathcal{R}_{\rightarrow}\}$$

flux mode: $f \in C$

EFM: $e \in C$ with *minimal support*

Elementary vectors in metabolic pathway analysis

- Elementary flux modes (EFMs) flux cone
Elementary flux vectors (EFVs) flux polyhedron (FBA)
- Elementary conversion modes (ECMs) exchange fluxes
- Elementary growth modes (EGMs) next-generation models
 resource balance analysis (RBA)
- eCXs microbial communities
eCOs, eXFs

Elementary vectors

Linear subspace S :

elementary vector (EV): $e \in S$ with minimal support

Theorem (Rockafellar 1969)

Every $f \in S$ is a finite, conformal sum of EVs:

$$f = \sum_e e \quad \text{with} \quad \text{sign}(e) \leq \text{sign}(f)$$

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Conformal refinement of Minkowski's theorem

Reaction directions, thermodynamics

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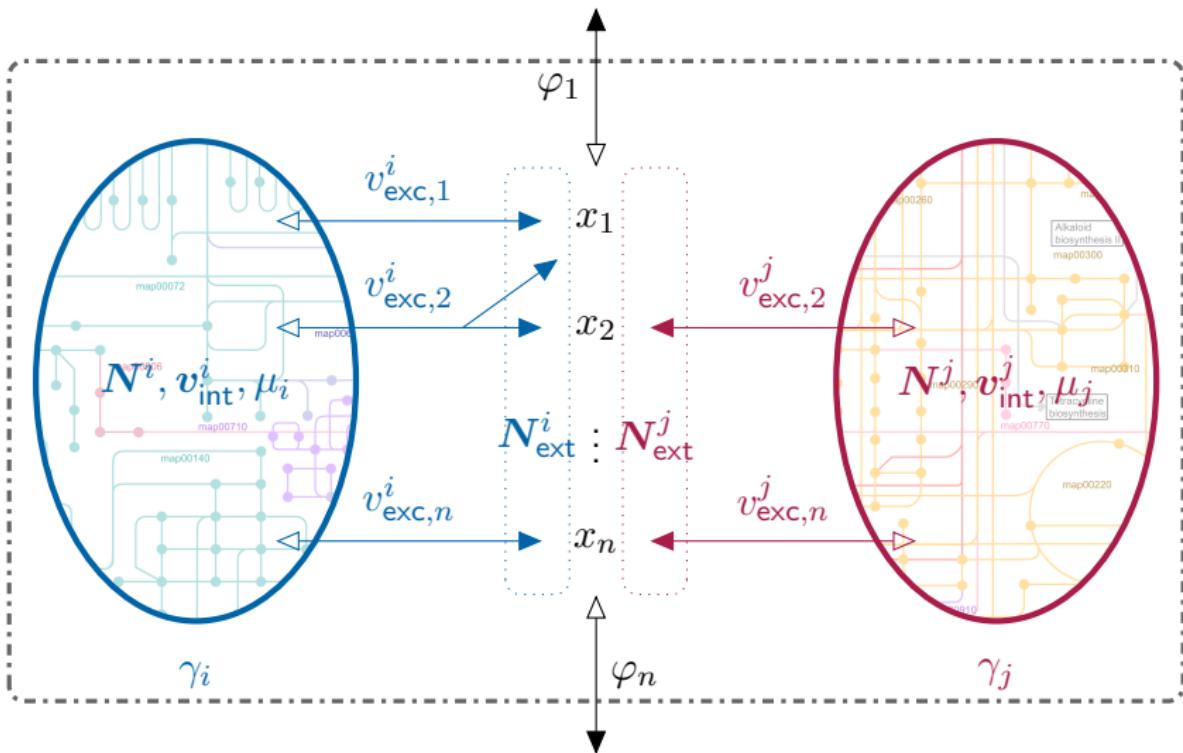
Reaction directions, thermodynamics

Generalization (Müller, Regensburger 2016)

linear subspace $S \rightarrow$ general polyhedral cone (e.g. flux cone C),
polyhedron (e.g. flux polyhedron P)

elementary vector: conformally non-decomposable

Microbial community



Single species: From ‘next-generation’ metabolic models ...

$$N_{\text{nxt-gen}} = \begin{matrix} & \text{RMet} & \text{RSyn} \\ \text{SMet} & N & S \\ \text{SMac} & 0 & I \end{matrix}$$

Growth dynamics:

$$\frac{dx}{dt} = N_{\text{nxt-gen}} v - \mu x$$

for $x = \frac{X}{M}$

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mass $M = \omega \cdot X$
vector of molar masses ω

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mass $M = \omega \cdot X$
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\implies

$$\mu = \omega \cdot N_{\text{nxt-gen}} v$$

... to 'standard' models

Steady state:

$$0 = \begin{pmatrix} N & S \\ 0 & I \end{pmatrix} \begin{pmatrix} v_{\text{Met}} \\ v_{\text{Syn}} \end{pmatrix} - \mu \begin{pmatrix} x_{\text{Met}} \\ x_{\text{Mac}} \end{pmatrix}$$

\implies

$$0 = Nv_{\text{Met}} - \mu \underbrace{(Sx_{\text{Mac}} + x_{\text{Met}})}_{\hat{x}_{\text{Met}}}$$

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Biomass “reaction”:

$$\sum_{i \in \text{Met}} \hat{x}_{\text{Met}} \cdot i \rightarrow \frac{1 \text{ mol}}{1 \text{ g}} \cdot \text{BM}$$

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\implies

$$0 = \underbrace{\begin{pmatrix} N & -n_{\text{bm}} \end{pmatrix}}_{N_{\text{bm}}} \begin{pmatrix} v_{\text{Met}} \\ v_{\text{bm}} \end{pmatrix}$$

$$n_{\text{bm}} = \hat{x}_{\text{Met}} \frac{\text{g}}{\text{mol}}, v_{\text{bm}} = \mu \frac{\text{mol}}{\text{g}}$$

Multiple species

$$0 = N_{\text{bm}}^i \left(\frac{v^i}{\mu_i \frac{\text{mol}}{\text{g}}} \right)$$

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Exchange metabolites (in medium):

$$\begin{pmatrix} ? \\ N^i \end{pmatrix} = \begin{matrix} \text{SExc} & \text{RExc} \\ \text{SMet} & \text{RInt} \end{matrix} \begin{pmatrix} N_{\text{ext}}^i & 0 \\ N_{\text{exc}}^i & N_{\text{int}}^i \end{pmatrix}$$

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Dynamics (in medium):

$$\frac{dX_{\text{SExc}}}{dt} = M \sum_i \gamma_i N_{\text{ext}}^i v_{\text{exc}}^i - \Phi$$

$$M = \sum_i M_i, \quad \gamma_i = \frac{M_i}{M}$$

Community model in μ , γ and v^i

$$0 = N_{\text{bm}}^i \left(\mu \frac{v^i}{\text{mol/g}} \right),$$
$$l^i \leq v^i \leq u^i,$$
$$\gamma_i \geq 0,$$

for $i = 1, \dots, \#\text{species}$, and

$$\sum_i \gamma_i (N_{\text{ext}}^i v_{\text{exc}}^i)_j \begin{cases} = 0, & \text{if } j \in \text{SExc}_0, \\ \leq 0, & \text{if } j \in \text{SExc}_{\text{in}}, \\ \geq 0, & \text{if } j \in \text{SExc}_{\text{out}}, \end{cases}$$
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Introduce $\bar{v}^i = \gamma_i v^i = \frac{v^i}{M}$ (fluxes on community level)

Koch, ..., Klamt (2019), Plos Comp. Biol.

Community model in μ , γ and \bar{v}^i

$$0 = N_{\text{bm}}^i \left(\frac{\bar{v}^i}{\gamma_i \mu \frac{\text{mol}}{\text{g}}} \right),$$
$$\gamma_i l^i \leq \bar{v}^i \leq \gamma_i u^i,$$
$$\gamma_i \geq 0,$$

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$$\sum_i \gamma_i = 1.$$

Project fluxes to exchange fluxes ($\bar{v}^i \rightarrow \bar{v}_{\text{exc}}^i$)

Community model in μ , γ and \bar{v}_{exc}^i

$$A^i \bar{v}_{\text{exc}}^i + b^i(\mu) \gamma_i \geq 0,$$

$$\begin{aligned} \gamma_i l_{\text{exc}}^i &\leq \bar{v}_{\text{exc}}^i \leq \gamma_i u_{\text{exc}}^i, \\ \gamma_i &\geq 0, \end{aligned}$$

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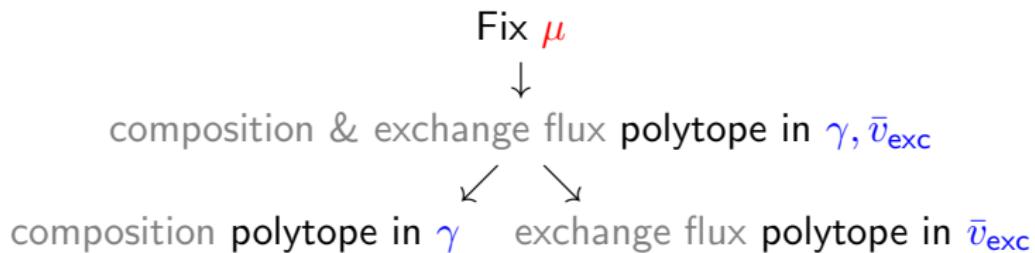
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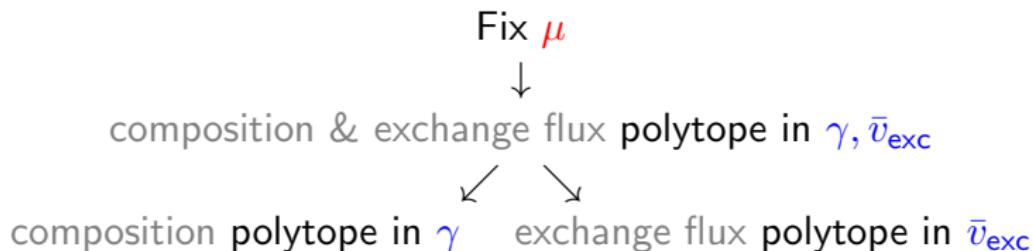
$$\sum_i \gamma_i = 1.$$

Fix μ (define polytope)

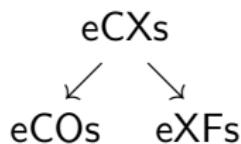
Community model \rightarrow community modes



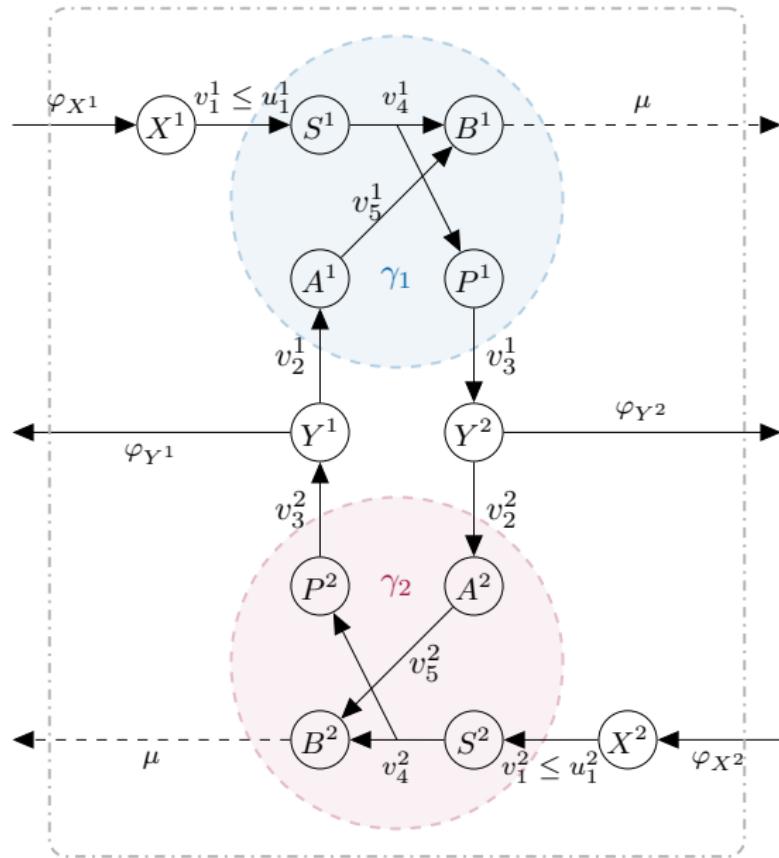
Community model \rightarrow community modes



Conformally non-decomposable vectors:



Example



Example: elementary compositions & exchange fluxes

$\hat{\mu} \in [0, 1]$:

$$\gamma_1, \gamma_2; \quad \hat{v}_1^1, \hat{v}_2^1, \hat{v}_3^1; \quad \hat{v}_1^2, \hat{v}_2^2, \hat{v}_3^2$$

$$eCX_1 = (1, 0; \hat{\mu}, 0, \hat{\mu}; 0, 0, 0)^\top,$$

$$eCX_2 = (0, 1; 0, 0, 0; \hat{\mu}, 0, \hat{\mu})^\top,$$

$$eCX_3 = (\frac{1}{2}, \frac{1}{2}; \frac{\hat{\mu}}{2}, 0, \frac{\hat{\mu}}{2}; 0, \frac{\hat{\mu}}{2}, 0)^\top,$$

$$eCX_4 = (\frac{1}{2}, \frac{1}{2}; 0, \frac{\hat{\mu}}{2}, 0; \frac{\hat{\mu}}{2}, 0, \frac{\hat{\mu}}{2})^\top.$$

Example: elementary compositions & exchange fluxes

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$\hat{\mu} \in [1, 2]$:

$$eCX_5 = (\quad \frac{1}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}}; \quad \frac{1}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}}, \frac{1}{\hat{\mu}}; \quad \frac{\hat{\mu}-1}{\hat{\mu}}, \frac{(\hat{\mu}-1)^2}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}} \quad)^\top,$$

$$eCX_6 = (\quad \frac{\hat{\mu}-1}{\hat{\mu}}, \frac{1}{\hat{\mu}}; \quad \frac{\hat{\mu}-1}{\hat{\mu}}, \frac{(\hat{\mu}-1)^2}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}}; \quad \frac{1}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}}, \frac{1}{\hat{\mu}} \quad)^\top,$$

$$eCX_7 = (\quad \frac{1}{2}, \frac{1}{2}; \quad \frac{1}{2}, \frac{\hat{\mu}-1}{2}, \frac{1}{2}; \quad \frac{\hat{\mu}-1}{2}, \frac{1}{2}, \frac{\hat{\mu}-1}{2} \quad)^\top,$$

$$eCX_8 = (\quad \frac{1}{2}, \frac{1}{2}; \quad \frac{\hat{\mu}-1}{2}, \frac{1}{2}, \frac{\hat{\mu}-1}{2}; \quad \frac{1}{2}, \frac{\hat{\mu}-1}{2}, \frac{1}{2} \quad)^\top.$$

Example: elementary compositions

$\hat{\mu} \in [0, 1]$:

$$\begin{aligned}\gamma_1, \gamma_2 \\ \text{eCO}_1 &= (1, 0)^\top, \\ \text{eCO}_2 &= (0, 1)^\top.\end{aligned}$$

$\hat{\mu} \in [1, 2]$:

$$\begin{aligned}\text{eCO}_5 &= (\frac{1}{\hat{\mu}}, \frac{\hat{\mu}-1}{\hat{\mu}})^\top, \\ \text{eCO}_6 &= (\frac{\hat{\mu}-1}{\hat{\mu}}, \frac{1}{\hat{\mu}})^\top.\end{aligned}$$

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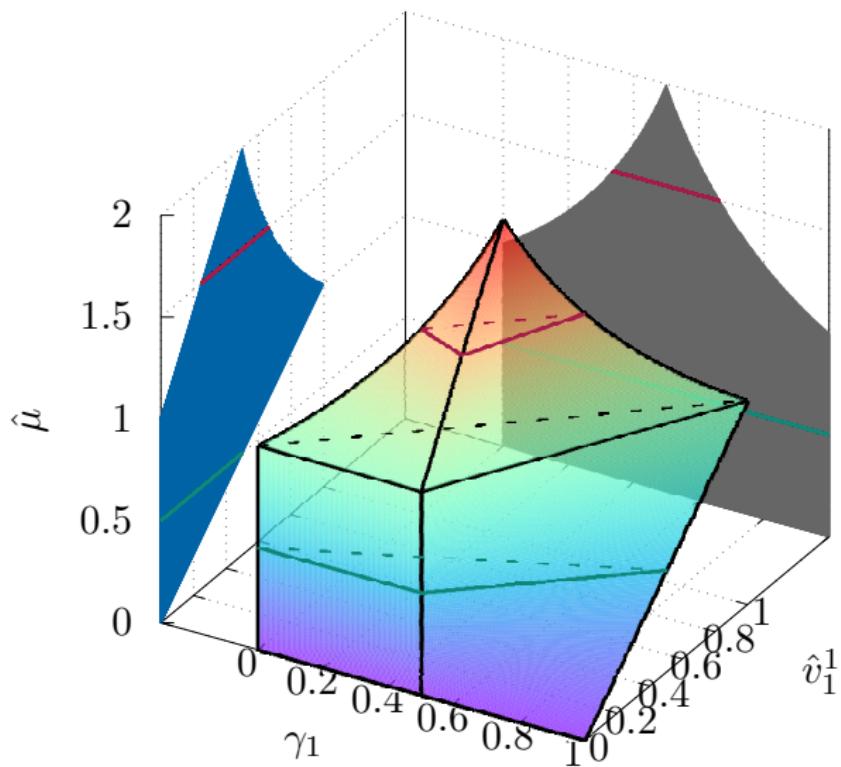
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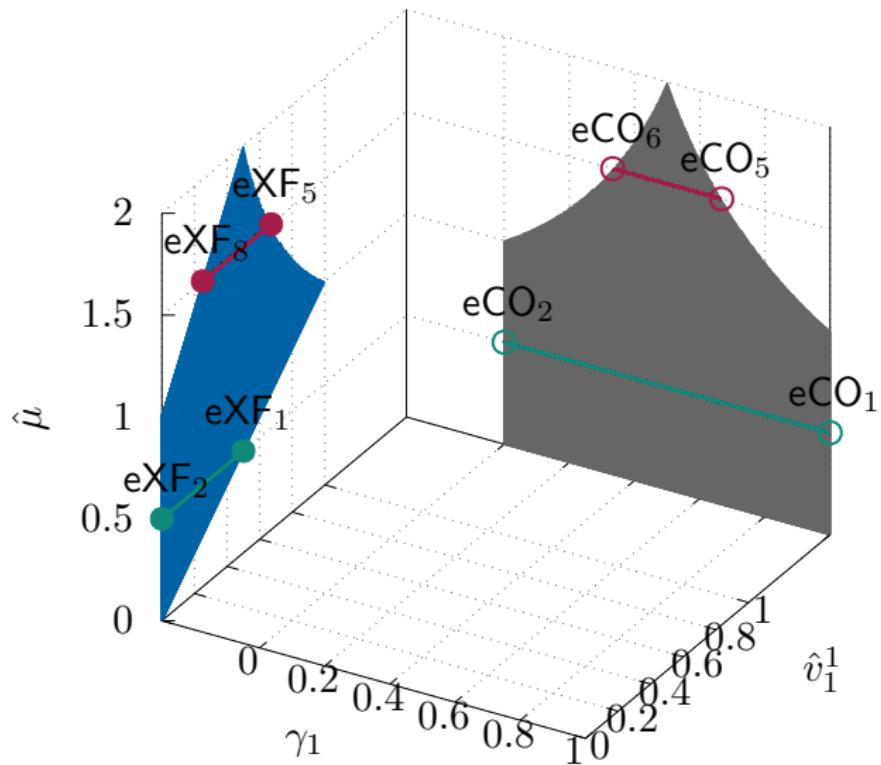
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analogously for elementary exchange fluxes (eXFs)

Example: projection to μ, γ_1, v_1^1



Example: eCOs and eXFs



Biological interpretation

eCX	$\hat{\mu} \in [0, 1]$				$\hat{\mu} \in (1, 2]$			
	1	2	3	4	5	6	7	8
specialization	✓	✓						
commensalism			✓	✓				
mutualism					✓	✓	✓	✓
maximum uptake					✓	✓		
maximum yield			✓	✓		✓	✓	
nonlinear in $\hat{\mu}$					✓	✓		

Next steps

- Applications:
 - small communities (with small/large metabolic models)
 - ecology?
- Computation:
 - projection, EV enumeration
 - DD, lrs

Advertisements

- Metabolic Pathway Analysis (MPA) 2025 Vienna
- Economic Principles in Cell Biology
Online seminar
A collaborative open-access textbook
<https://principlescellphysiology.org>
- Mathematics of Reaction Networks (MoRN)
Online seminar
<https://researchseminars.org/seminar/MoRN>

Thank you for your attention!