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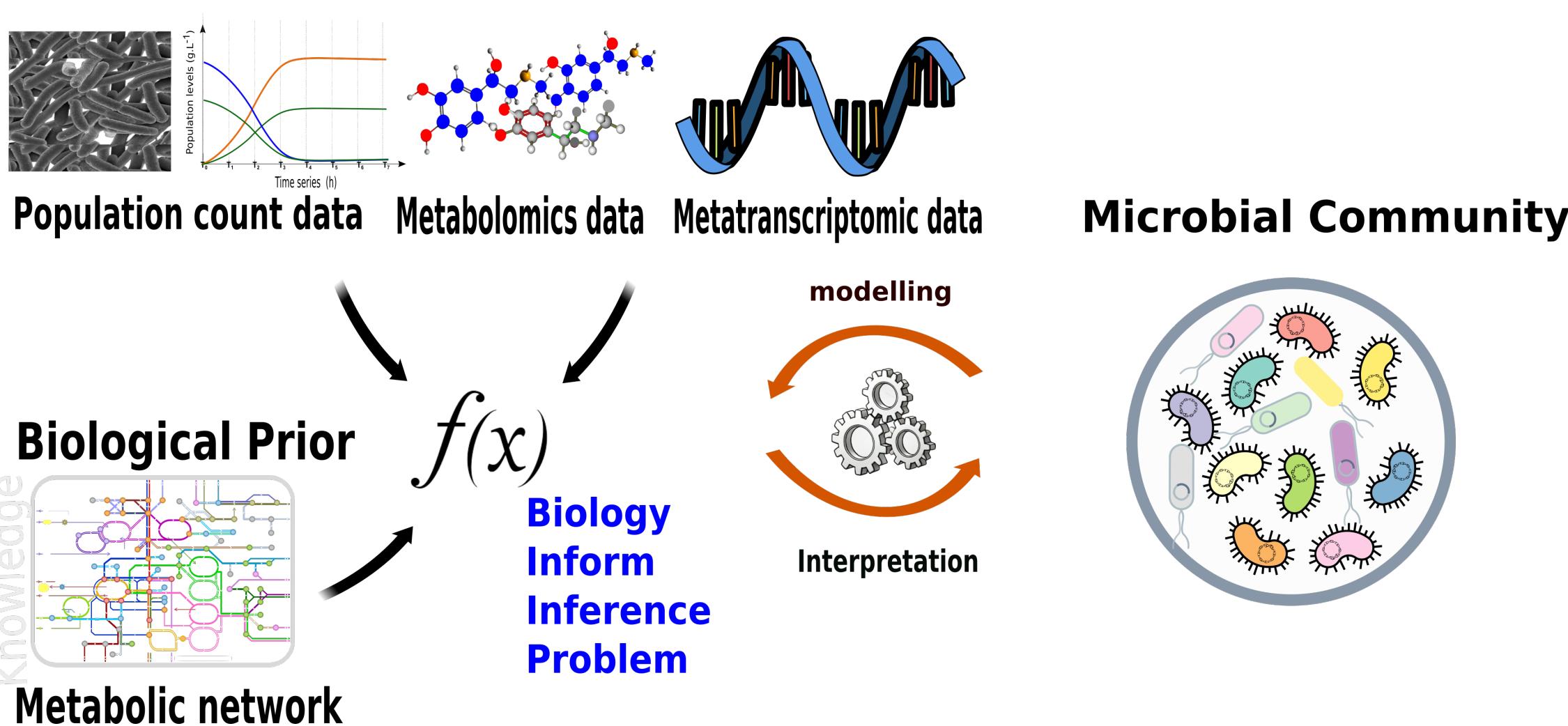
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cMFA for multi-omics data integration in microbial community models

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INTRODUCTION

Biology-informed inference from multi-omics data



Understanding natural microbial communities presents significant challenges due to the complex interactions and metabolite exchanges between microorganisms. To gain deeper insights into these dynamics, we are focusing on:

- ⌚ Developing mathematical models based on metabolic networks.
- ⌚ Inferring metabolic fluxes using biology-informed high-dimensional analysis methods to integrate multi-omics data.

Scientific Question:

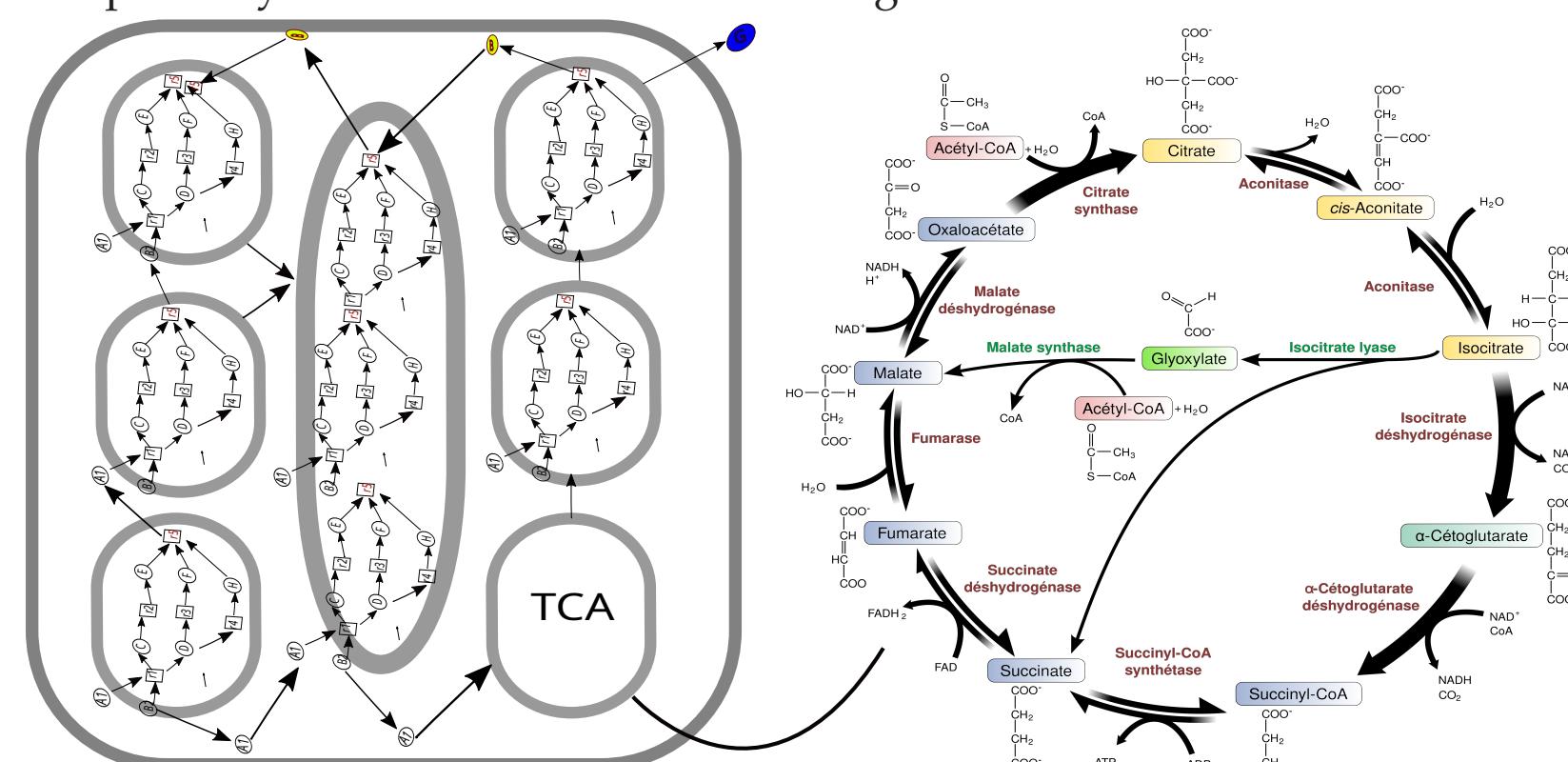
How can mathematical models help us better understand the functioning of microbial communities?

- ⌚ Internal flux reconstruction based on some external reaction rates.
- ⌚ Estimating the contribution of each microorganism to the community.

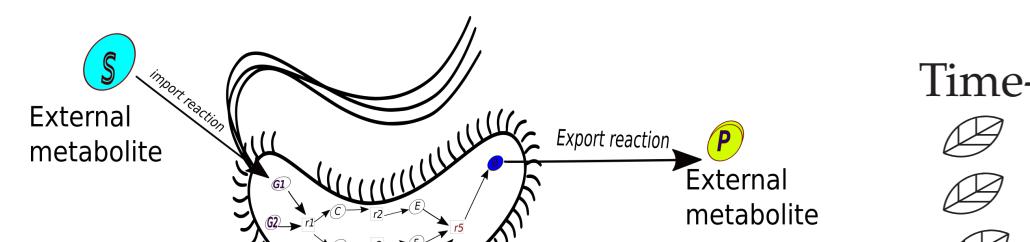
COMMUNITY METABOLIC FLUX ANALYSIS (cMFA)

METABOLIC NETWORK

A metabolic network refers to the interconnected set of biochemical reactions and pathways that occur within a cell or organism.



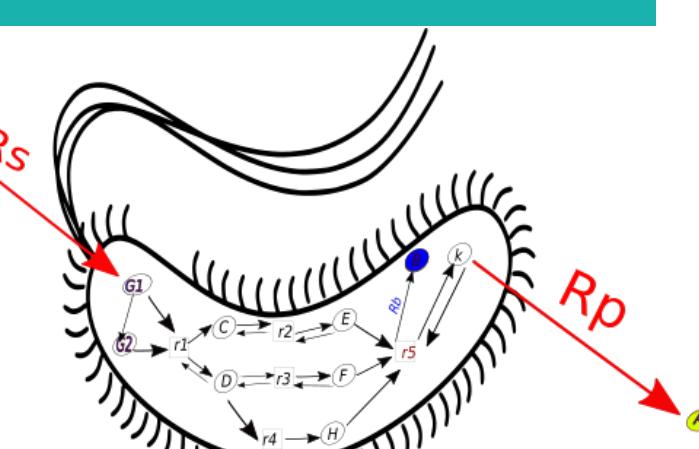
INPUT DATA DESCRIPTION



Time-series of:
⌚ Targeted metabolomic data.
⌚ Microbial counts.
⌚ Metatranscriptomic (expressed metabolic reactions).

METABOLIC FLUX ANALYSIS (MFA)

Experimental technique used to estimate intracellular metabolic fluxes by analyzing metabolite production and consumption rates in a biological system, providing insights into the cell's core metabolism



COMMUNITY MFA (cMFA)

Find, for $t \geq 0$, for a set \mathcal{M} of N_m metabolites and for a set \mathcal{B}_m of N_b micro-organisms the flux vector

$$\hat{\nu}(t) = (\nu^{(b_1)}(t), \dots, \nu^{(b_{N_b})}(t)) \text{ of dimension } N_r \text{ such that}$$

$$\hat{\nu}(t) := \arg \min \sum_{m \in \mathcal{M}} \left\| R_{m,t} - \sum_{b \in \mathcal{B}_m} x_t^{(b)} \nu_{m,t}^{(b)} \right\|_2^2 + \lambda \mathcal{R}(\nu)$$

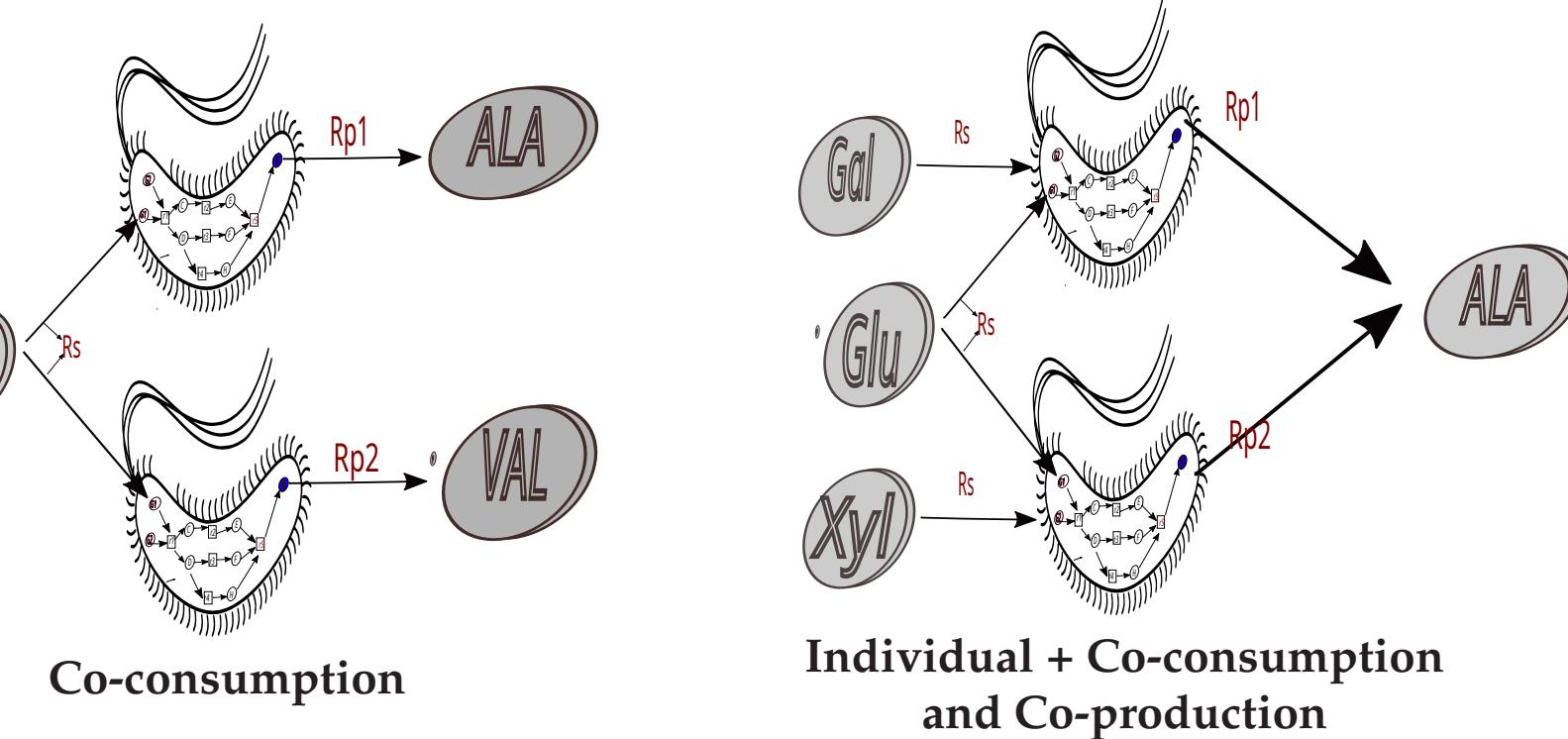
$$S.T : \begin{cases} S^{(b)} \cdot \nu^{(b)} = 0, \text{ for } b \in \mathcal{B}_m \\ C_{min}^{(b)} \leq \nu^{(b)} \leq C_{max}^{(b)}, \text{ for } b \in \mathcal{B}_m \end{cases}$$

Regression problem is:

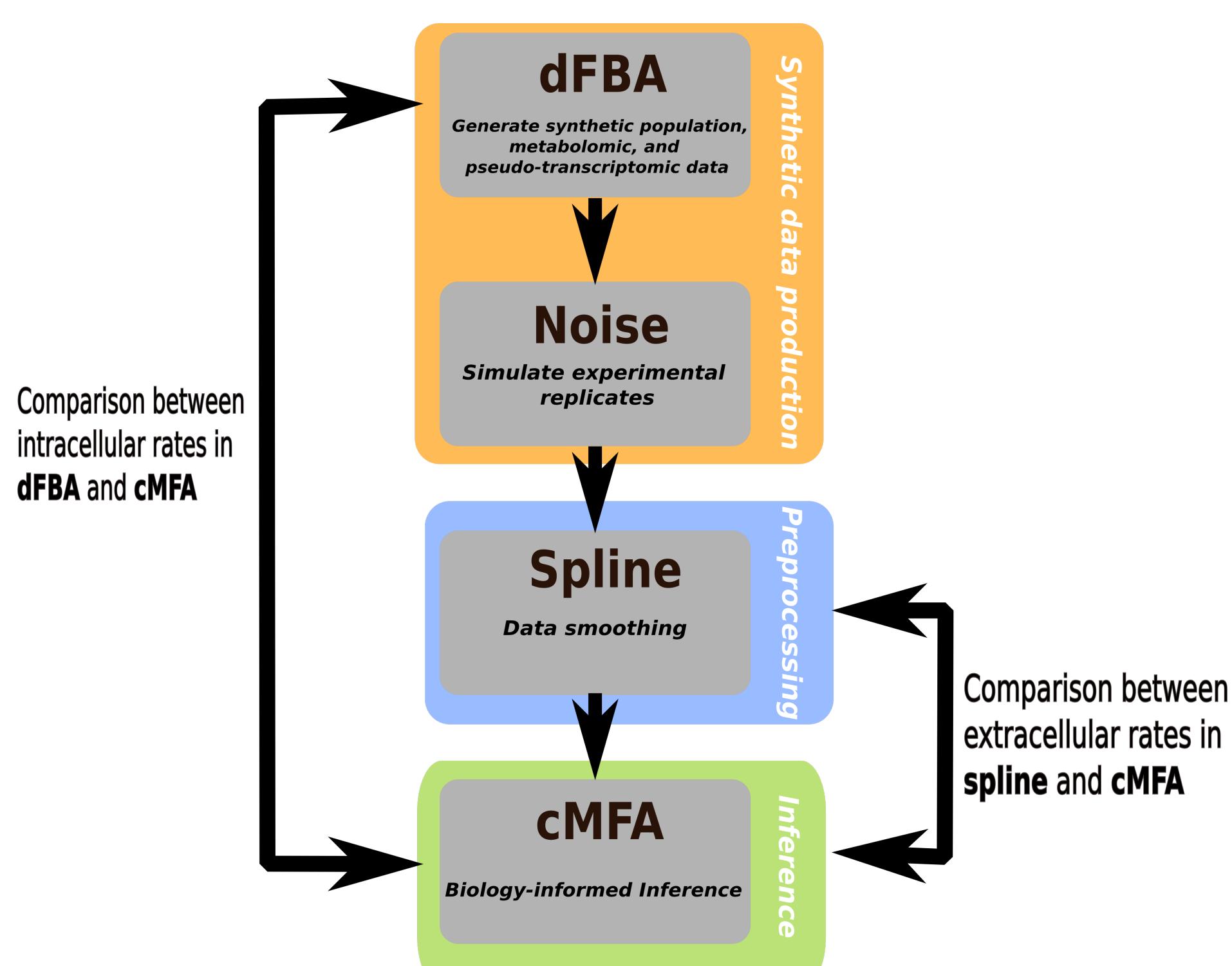
- ⌚ High-dimensional ($N_r \sim 5000$)
- ⌚ Highly-underdetermined ($N_r \gg N_m$)
- ⌚ Biology-informed ($S^{(b)}$)
- ⌚ Constrained by metatranscriptomics (where $C_{min}^{(b)} = C_{max}^{(b)} = 0$)
- ⌚ Regularized by ridge ($\mathcal{R}(\nu) = \|\nu\|_2^2$) or lasso ($\mathcal{R}(\nu) = \|\nu\|_1$) penalty

CO-CULTURE DEFINITION

Biological situation of E. coli to produce amino acids and ALA super producer



BENCHMARK METHODOLOGY



GENERATING DATA: dFBA

Dynamic Flux Balance Analysis (DFBA) is an extension of traditional Flux Balance Analysis (FBA), which is a mathematical modeling approach used in systems biology to predict the flow of metabolites through metabolic networks.

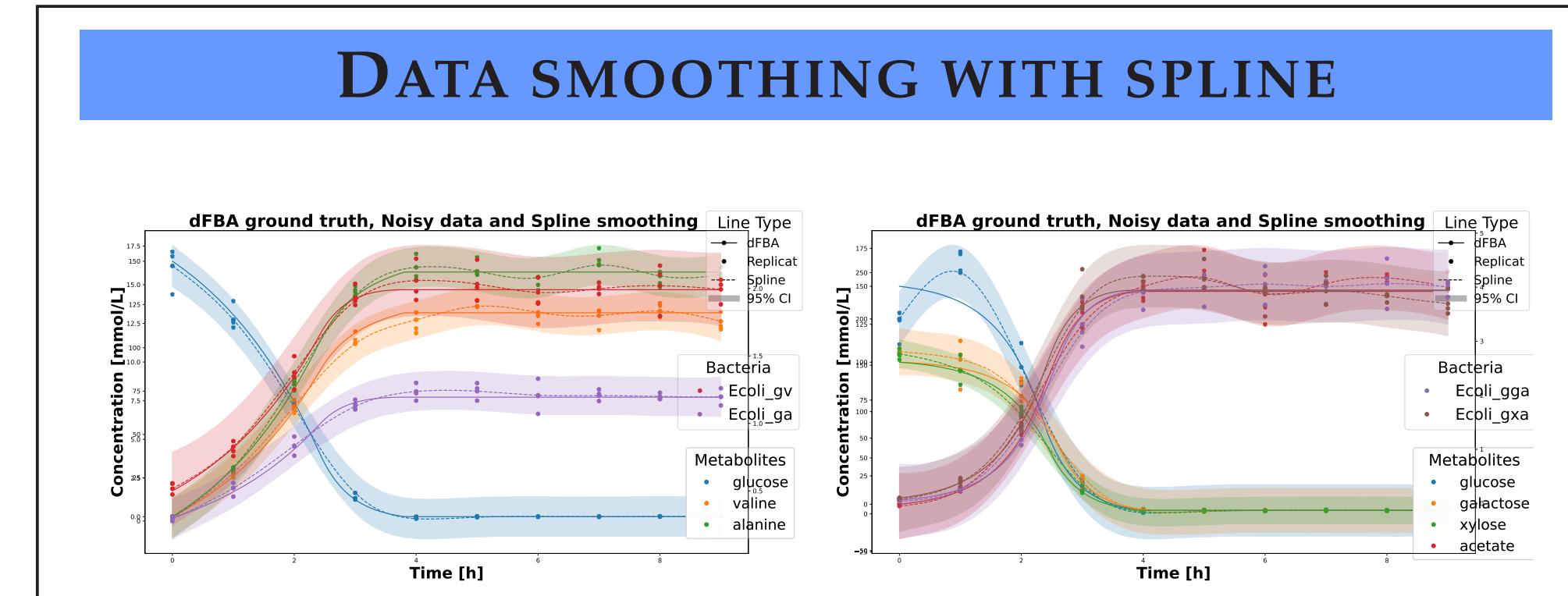
$$\begin{aligned} \partial_t b_i &= \sum_{j \in \mathcal{C}} \hat{\nu}_{i,j}(M, b) b_j \\ \hat{\nu}_i &= \arg \max B \nu_i \\ S.T \quad &\begin{cases} S_i \hat{\nu}_i = 0 \\ C_M \leq \hat{\nu}_i \leq C_{max} \\ 0 \leq \hat{\nu}_i \leq C_{max} \\ -C_{max} \leq \hat{\nu}_i \leq C_{max} \end{cases} \end{aligned}$$

IMPLEMENTATION

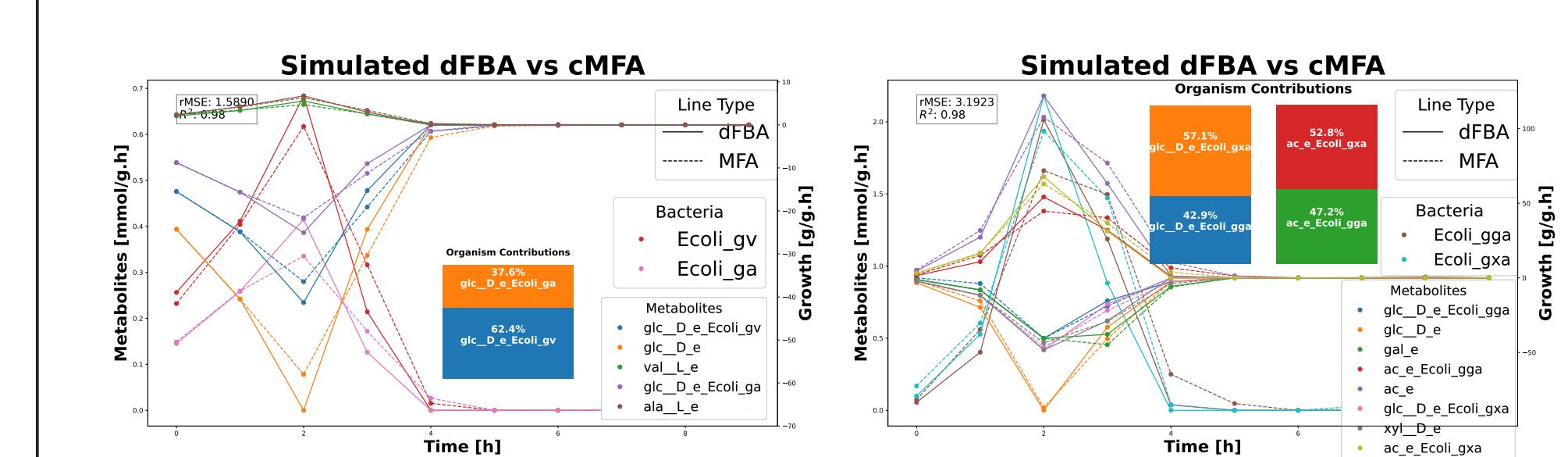
- NumPy and Pandas for data handling and structure.
- PyGAM for spline smoothing.
- Matplotlib for visual representations.
- COBRAPy for constraint-based modeling (dFBA).
- OSQP for quadratic programming problems.

RESULTS

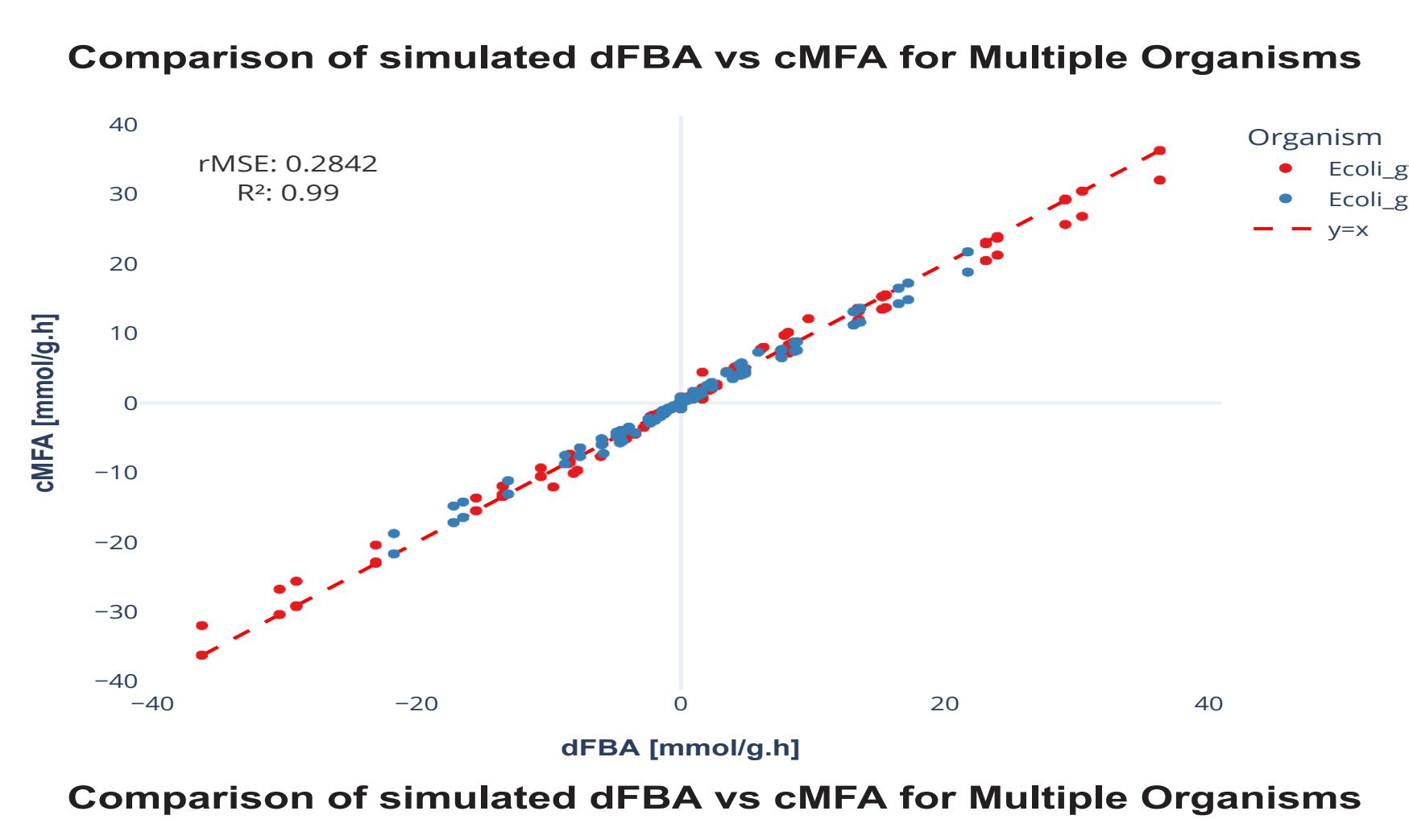
DATA SMOOTHING WITH SPLINE



INFERRING EXCHANGE FLUXES: DFBA VS cMFA



INFERRING INTERNAL FLUXES



PERSPECTIVES

- Benchmark on real data.
- Expand the community model to incorporate a diverse set of more than 10 to 20 microorganisms.
- The extension of MFA and the inclusion of metatranscriptomic data

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