

Reasoning approaches for the characterization of cooperation and competition in large-scale microbial communities

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Reasoning approaches for the characterization of cooperation and competition in large-scale microbial communities

Maxime Lecomte – David Sherman – Hélène Falentin – Clémence Frioux

GT-BIOSS NANTES November, 17 2022



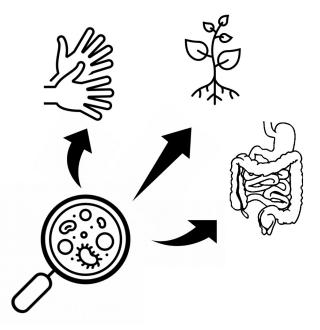


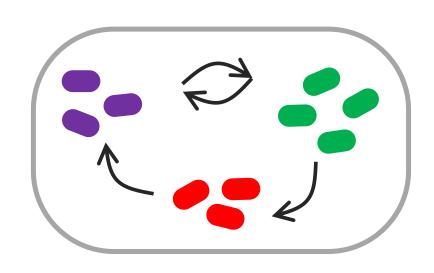
Motivation

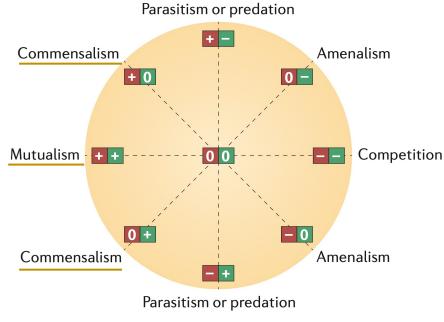
Bacterial communities?

System biology

Interaction types







Microbes are essential in ecosystems

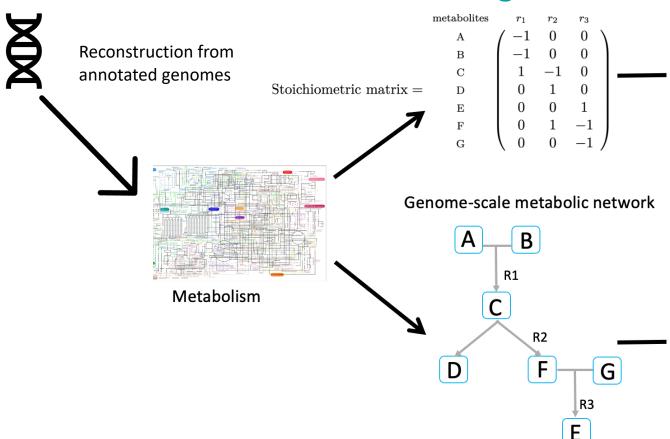
Bacterial interactions as a key mechanism

Cooperation and competition

Metabolism enable the understanding of interaction-based mechanisms in microbial communities



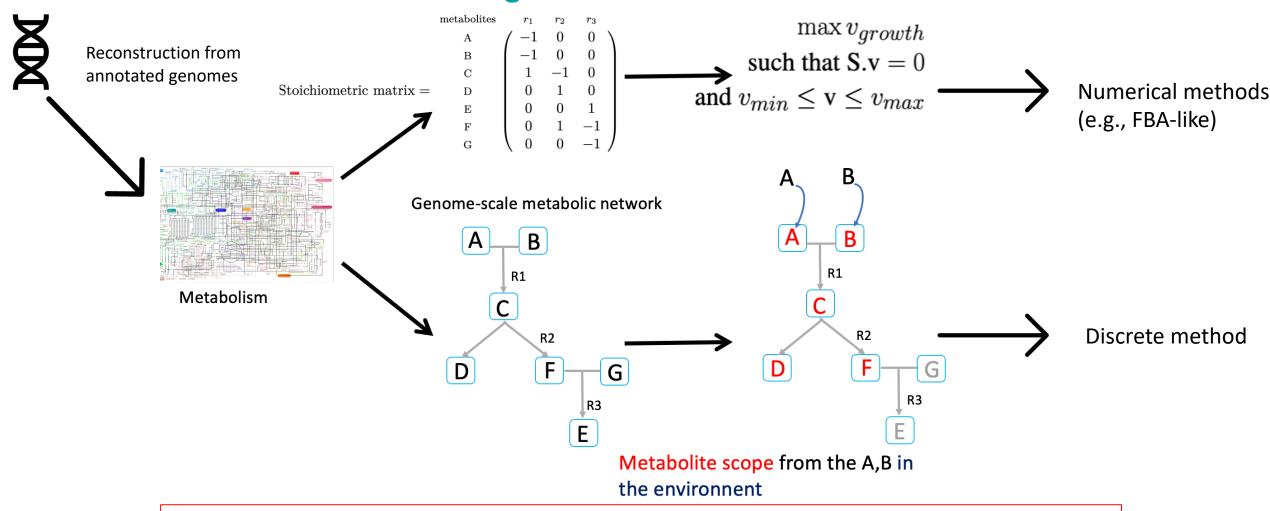
> State-of-the art for modeling bacterial communities



Genome-scale metabolic networks describe all functions associated to genomes



State-of-the art for modeling bacterial communities



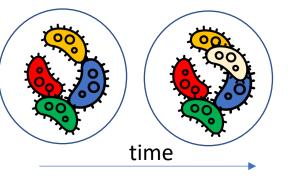
Two mathematical formalisms used for modeling the behavior of an organism in its environment



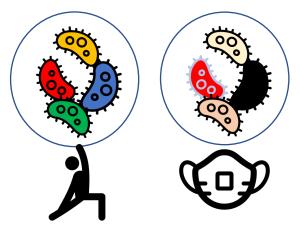
➤ Goal : Identify cooperation and competition potential in large scale bacterial communities

Cooperation and competition potential

Longitudinal analysis



Cross sectional analysis



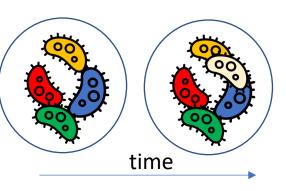


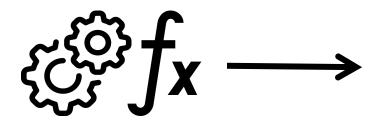
➤ Goal: Identify cooperation and competition potential in large scale bacterial communities

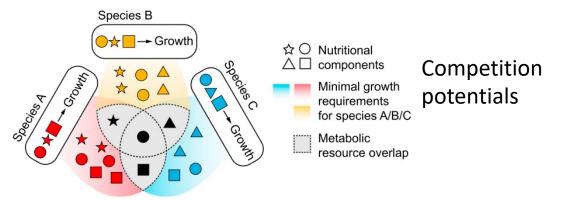
Cooperation and competition potential

Nutrient point of view

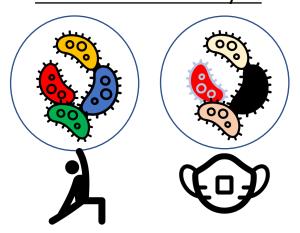
Longitudinal analysis







Cross sectional analysis



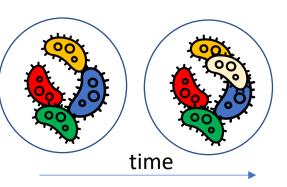


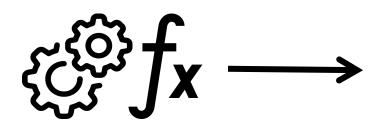
➤ Goal : Identify cooperation and competition potential in large scale bacterial communities

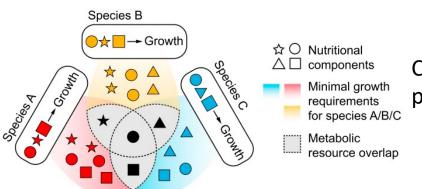
Cooperation and competition potential

Nutrient point of view

Longitudinal analysis

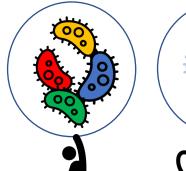


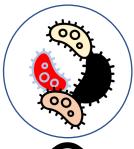




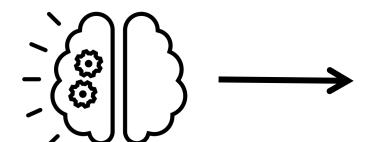
Competition potentials

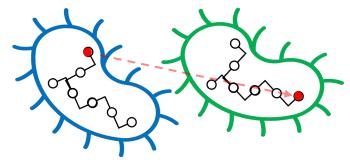
Cross sectional analysis











Cooperation potentials

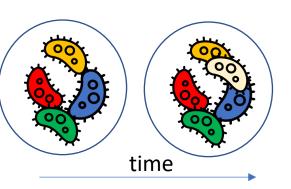
Either numerical or discrete methods highlight cooperation and competition potentials

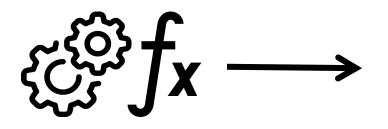
INRAO

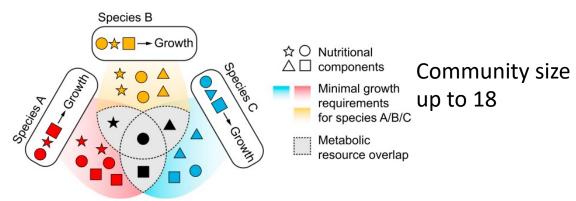
➤ Goal: Identify cooperation and competition potential in large scale bacterial communities

Scale-up to large scale bacterial communities

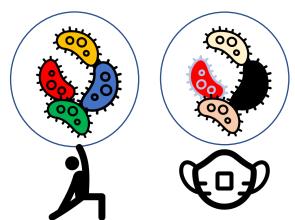
Longitudinal analysis







Cross sectional analysis



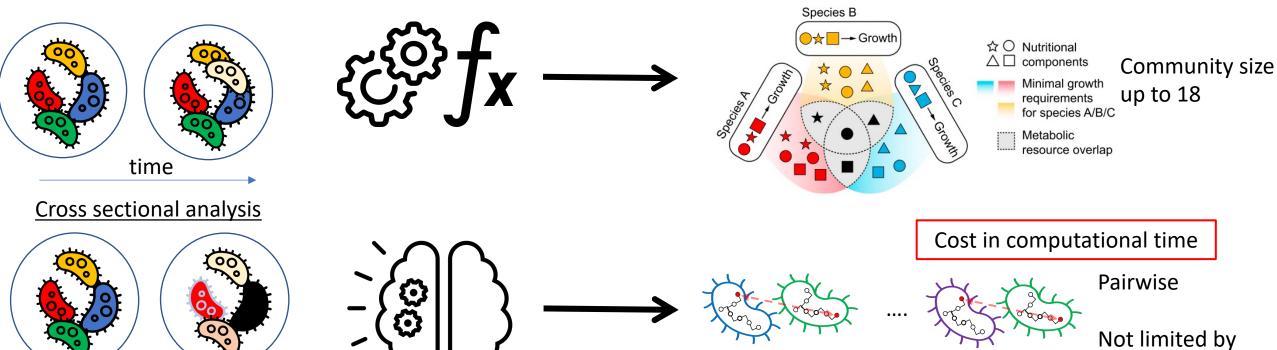
Cost in computational time

INRAe

Goal: Identify cooperation and competition potential in large scale bacterial communities

Scale-up to large scale bacterial communities

Longitudinal analysis



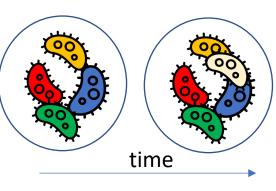
INRAe

size

the community

Avoid pairwise analysis for characterizing cooperation and competition in microbial communities

Longitudinal analysis



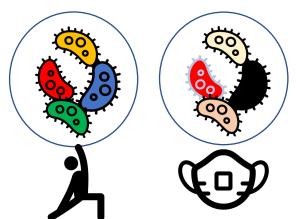
Characterize community



Pairwise analysis

(150 * (150-1))/2 = 11 375 combinations

Cross sectional analysis

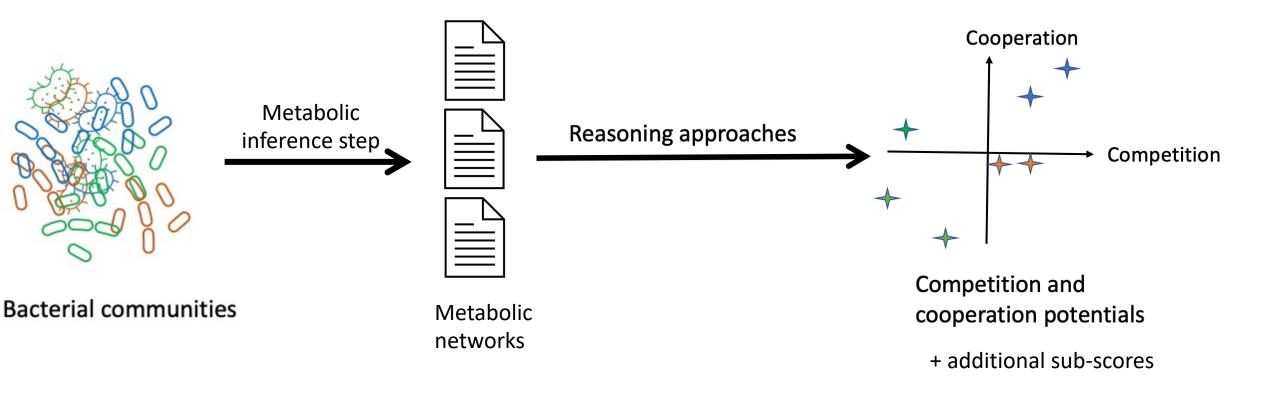


Cost in computational time + tedious analysis

How to characterize cooperation and competition potentials? → trade-off between scalability and accuracy

INRAO

Avoid pairwise analysis for characterizing cooperation and competition in microbial communities



Calculation of scores for characterizing the whole community with Answer Set Programming



Answer Set Programming

Logic paradigm and Knowledge representation & reasoning

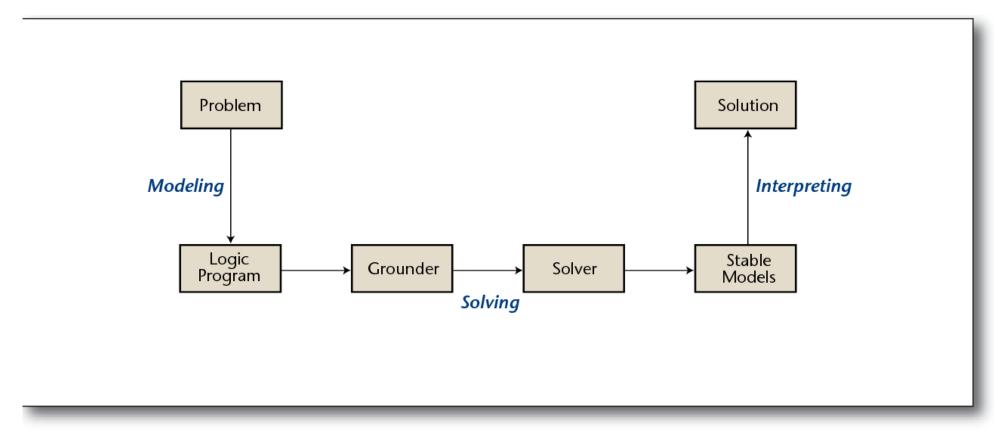


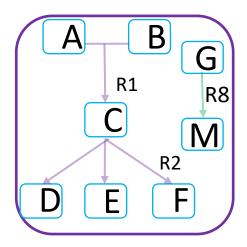
Figure 1. The Work Flow of Answer Set Programming.

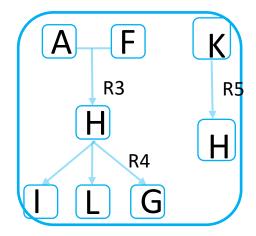


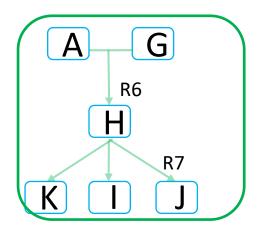
Kaufmann, Benjamin et al. "Grounding and Solving in Answer Set Programming." Al Mag. 37 (2016): 25-32.

Cooperation potential based on exchanged metabolites using ASP

Seed: A,B



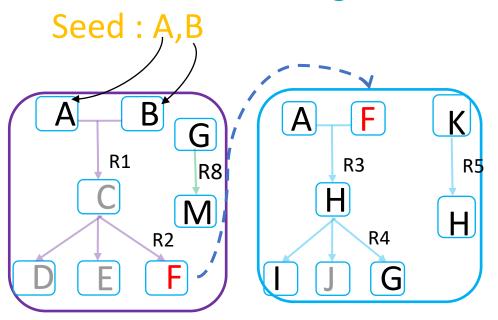


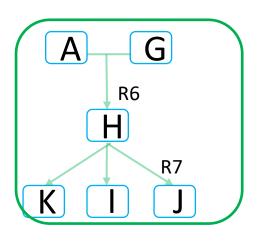


Metabolites

 $exchanged_{metabolites} = scope(G_S, S), not indivProducible(S, G_t), products(S, G_S).$ Where $S \in G$ and $T \in G$, and $S \neq T$.

Cooperation potential based on exchanged metabolites using ASP





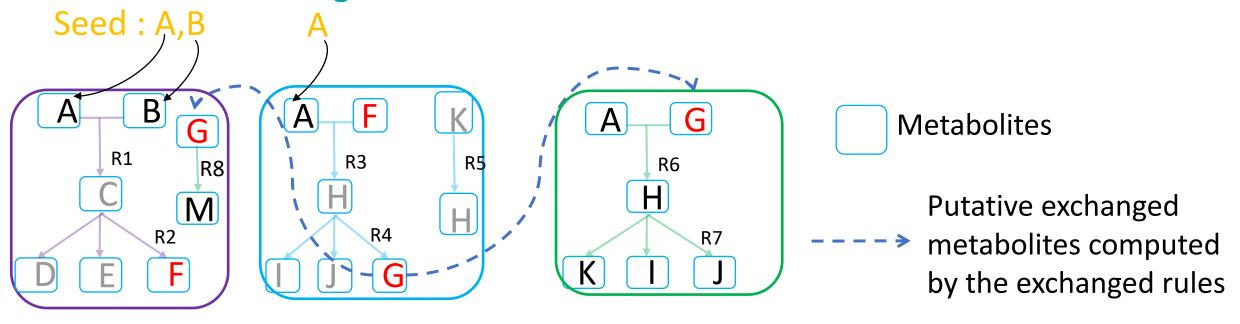
Metabolites

Putative exchangedmetabolites computedby the exchanged rules

From	What	To
	F	



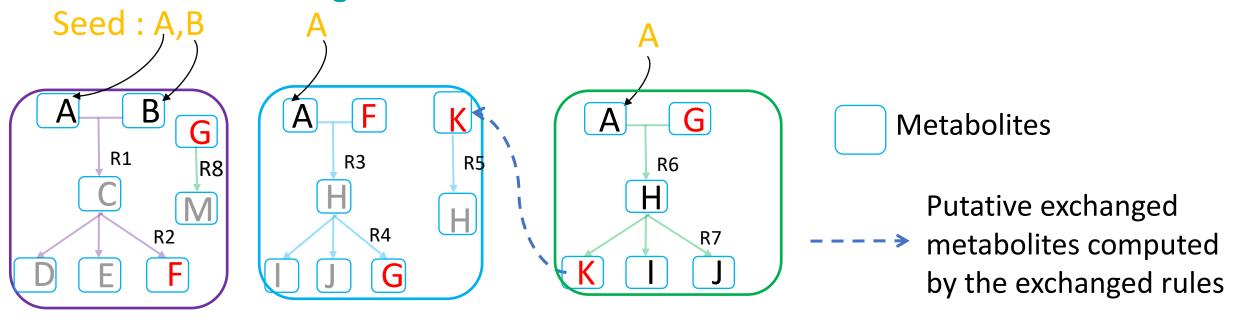
Cooperation potential based on exchanged metabolites using ASP



From	What	<u>To</u>
	F	
	G	
	G	

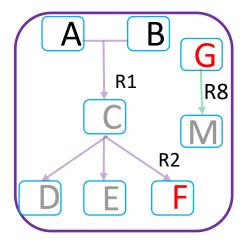


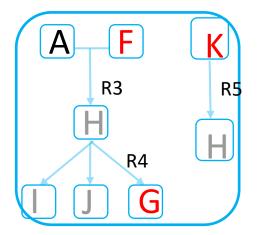
Cooperation potential based on exchanged metabolites using ASP

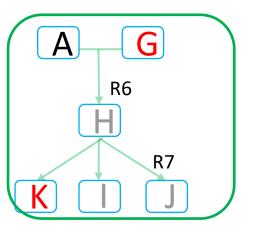


From	What	To
	F	
	G	
	G	
	K	
INRAE		

hypothesis: each species contributes differently

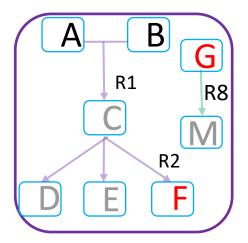


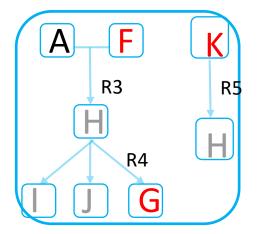


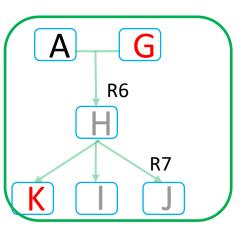


From	What	To
	F	
	G	
	G	
	K	
INRAE		

hypothesis: each species contributes differently







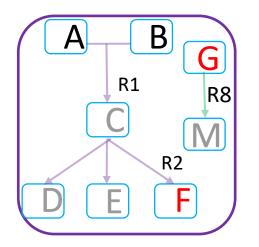
<u>From</u>	What	<u>To</u>
	F	
	G	
	G	
	K	
INRAE		

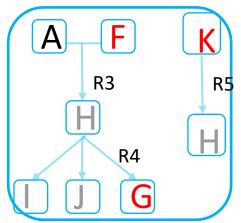
$$cooperation = \sum_{i=1}^{m} \sum_{k=2}^{n} 1 + 2^{-(k-1)}$$

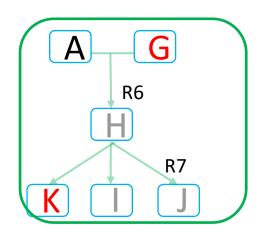
Number of metabolites

Number of different producers / consumers

hypothesis: each species contributes differently







Metabolites	Producers	consumers
F	1	1

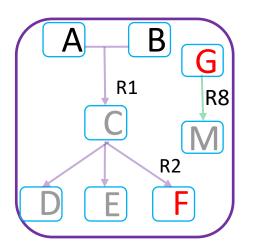
<u>From</u>	What	<u>To</u>
	F	
	G	
	G	
	K	
INRAE		

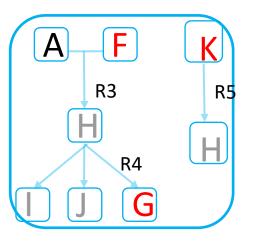
$$cooperation = \sum_{i=1}^{m} \sum_{k=2}^{n} 1 + 2^{-(k-1)}$$

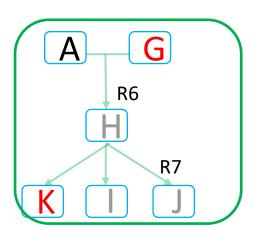
Number of metabolites

Number of different producers / consumers

hypothesis: each species contributes differently







Metabolites	Producers	consumers
F	1	1
G	1	1 + 2 ¹ =1.5

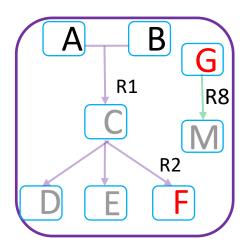
<u>From</u>	What	<u>To</u>
	F	
	G	
	G	
	K	
INRAE		

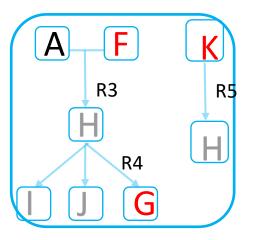
$$cooperation = \sum_{i=1}^{m} \sum_{k=2}^{n} 1 + 2^{-(k-1)}$$

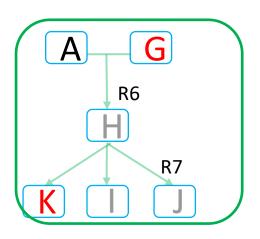
Number of metabolites

Number of different producers / consumers

hypothesis: each species contributes differently

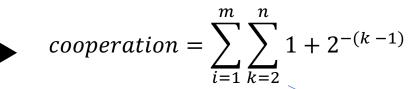






Metabolites	Producers	consumers
F	1	1
G	1	$1 + 2^1 = 1.5$
K	1	1

Cooperation = F + G + K = 6.5

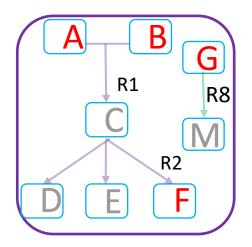


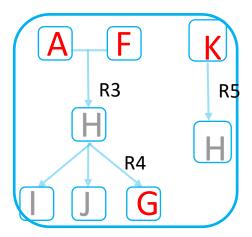
Number of metabolites

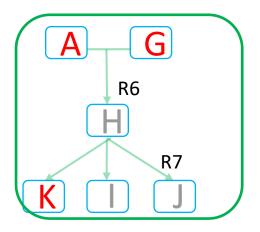
Number of different producers / consumers

Competition potential based on limiting substrate using ASP

Seed: A,B







G Limiting substrates

A,B,F

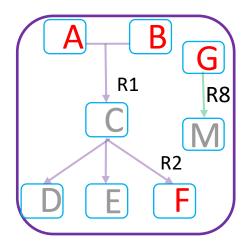
A,B,G,K

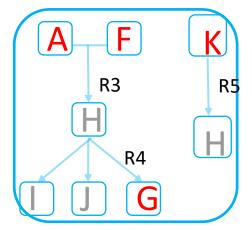
A,G,K

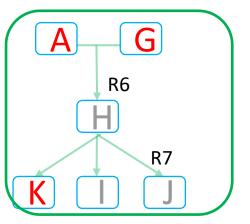


> Calculation of the competition potential (python)

Seed: A,B







$$competition = \sum_{i=1}^{m} \frac{count(i)}{size_community}$$

A,G,K

Calculation of the competition potential (python)

A:3

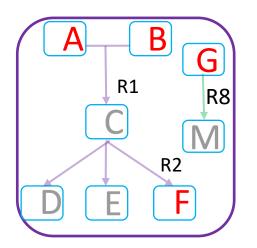
B:2

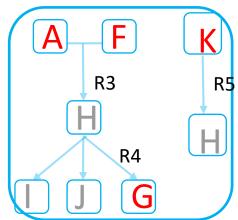
G:2

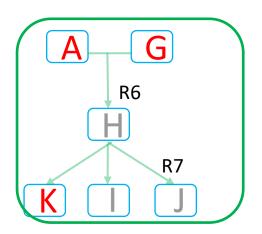
K:2

F:1

Seed: A,B







Metabolites Number of consumers

$$competition = \sum_{i=1}^{m} \frac{count(i)}{size_community}$$

INRAO

> Calculation of the competition potential (python)

A:3

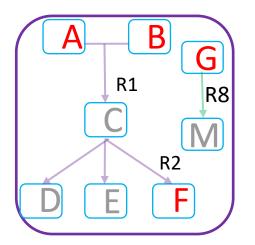
B: 2

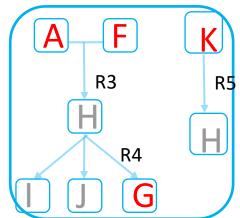
G:2

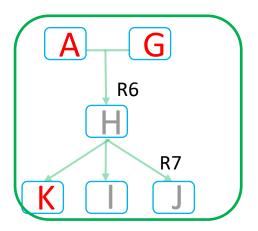
K:2

F:1

Seed: A,B







Metabolites

Number of consumers

A:1

B: 0.6

G: 0.6

K: 0.6

F: 0.3

Regulation by the size of the community (3 here)

$$competition = \sum_{i=1}^{m} \frac{count(i)}{size_community}$$



> Calculation of the competition potential (python)

A:3

B:2

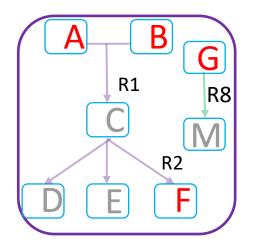
G:2

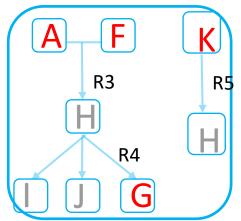
K:2

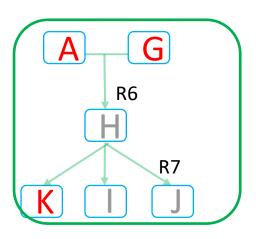
F:1

Competition = 1 + 0.6 + 0.6 + 0.6 + 0.3 = 3.1

Seed: A,B







Metabolites ,

Number of consumers

A:1

B: 0.6

G: 0.6

K: 0.6

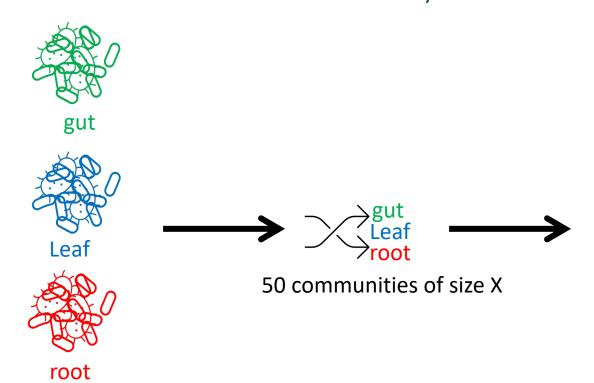
F: 0.3

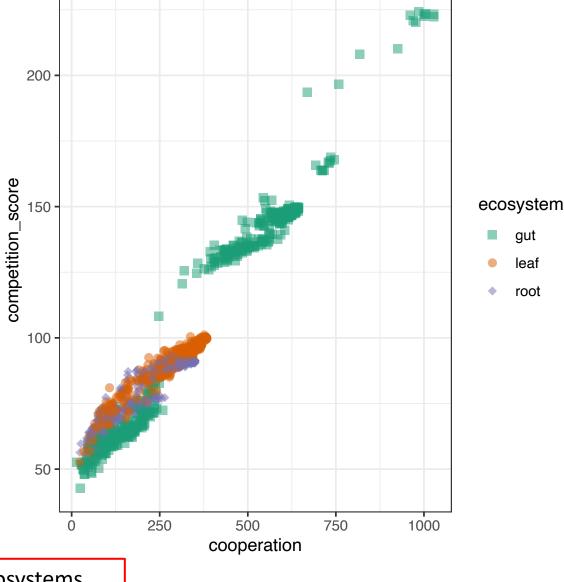
Regulation by the size of the community (3 here)

$$competition = \sum_{i=1}^{m} \frac{count(i)}{size_community}$$



Benchmarks for testing scores Scores in function of the ecosystem





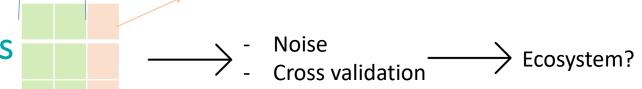
Reference genomes of cultivable species from 3 ecosystems

Different scores in function of the ecosystems



Ecosystem prediction from scores

Prediction of gut ecosystem (SVM)

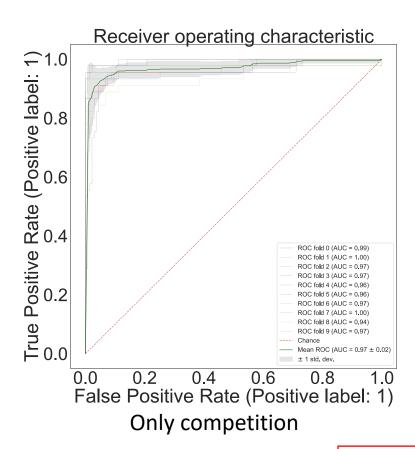


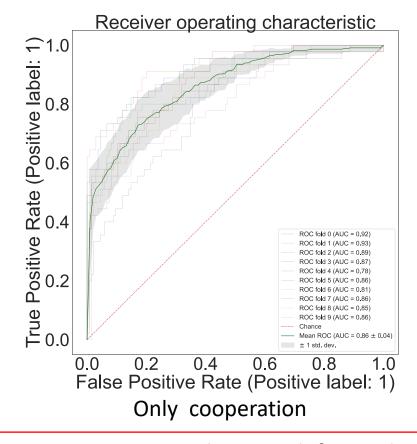
ecosystem

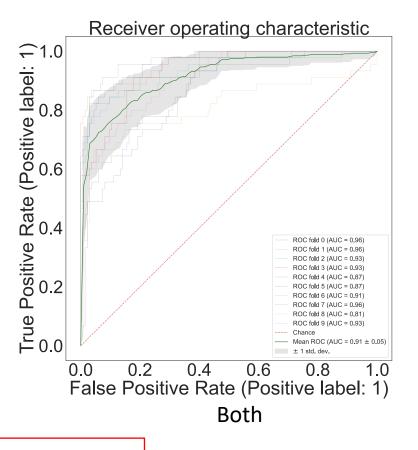
Up to 3 features = training

scores

Up to 4





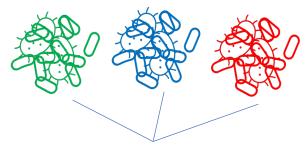


INRAe

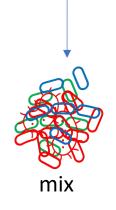
Competition score seems to be enough for predicting ecosystem

Benchmarks for testing scores

Do scores differ between ecosystems?



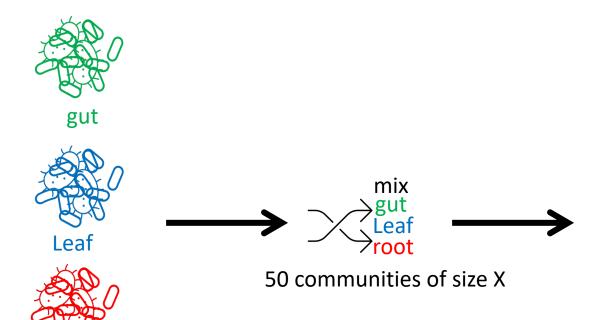
33% of each ecosystem

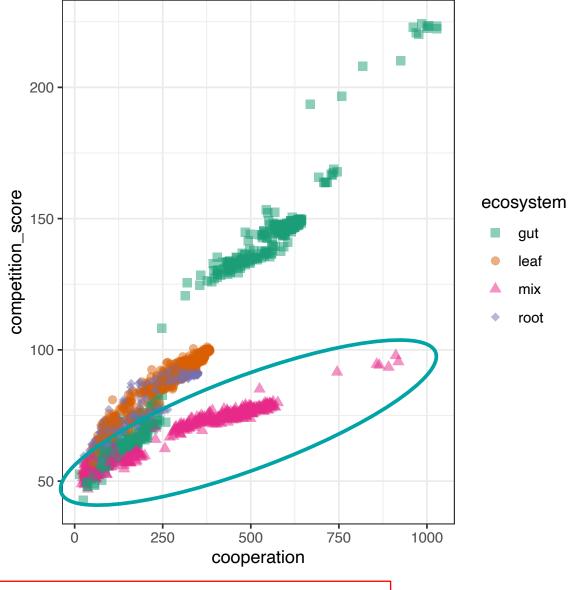


Cooperation and competition scores significantly differ between non-realistic and ecosystem-consistent communities



Benchmarks for testing scores Scores in function of the ecosystem



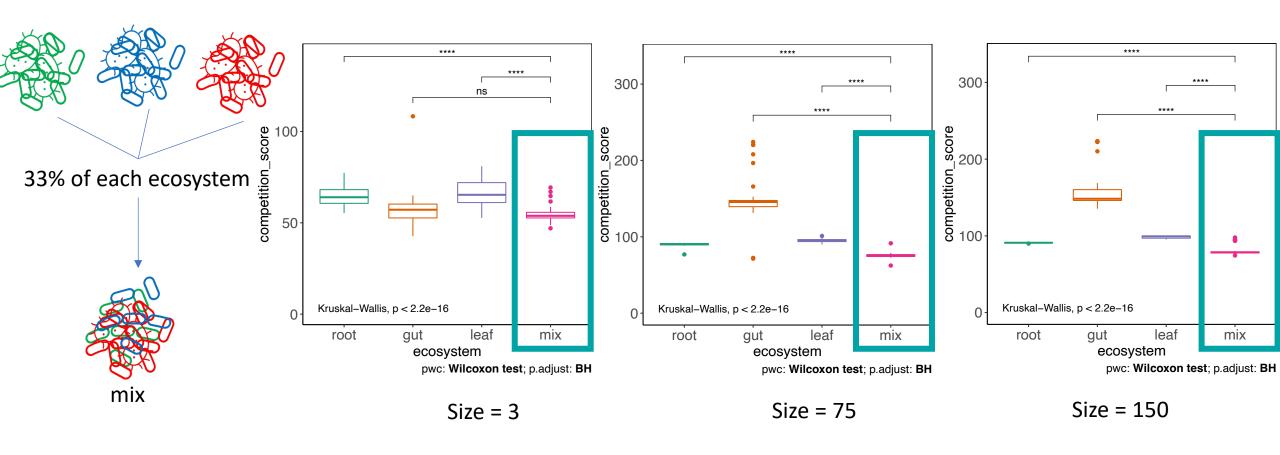




Cooperation and competition scores significantly differ between non-realistic and ecosystem-consistent communities

Benchmarks for testing scores

Do competition score differ between ecosystems?

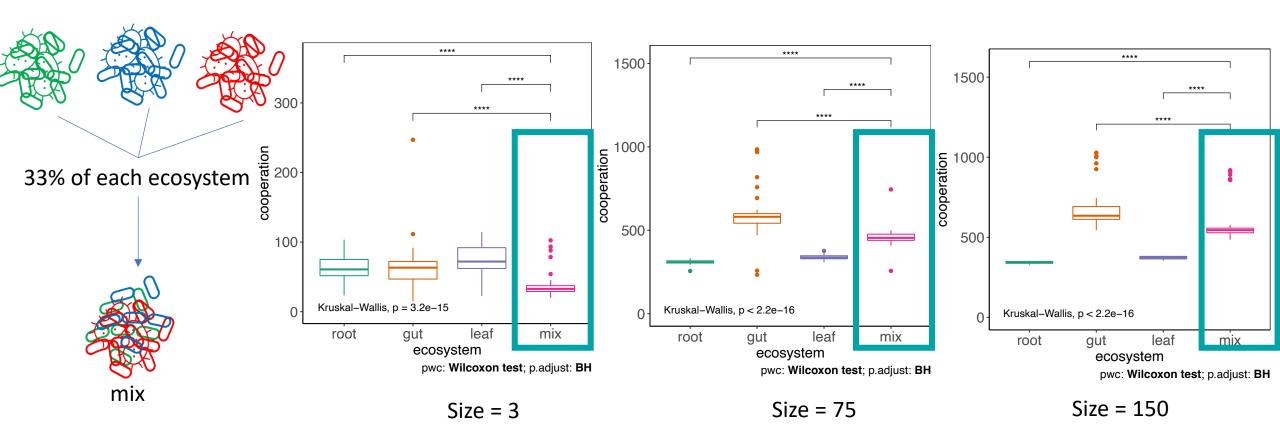


Cooperation and competition scores significantly differ between non-realistic and *ecosystem-consistent* communities



Benchmarks for testing scores

Do cooperation score differ between ecosystems?



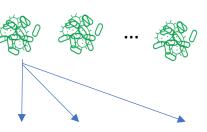
Cooperation and competition scores significantly differ between non-realistic and *ecosystem-consistent* communities



Benchmarks for testing scores Added-value of adding a bacteria in community



1520 isolated genomes of gut microbiome



50 random communities of size X

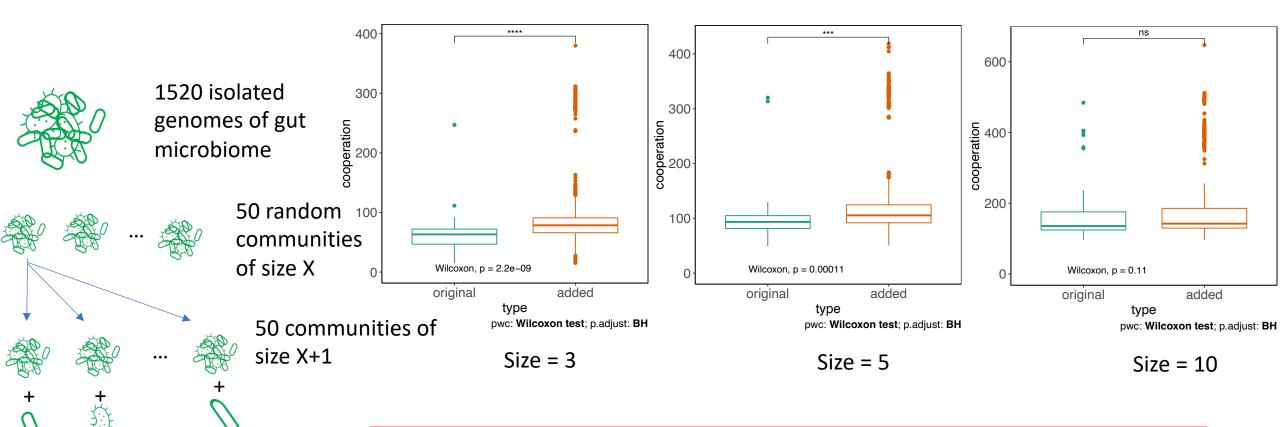


50 communities of size X+1



Benchmarks for testing scores

Added-value of adding a bacteria in community based on cooperation score



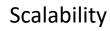
The larger is the community the less the community is disturbed by another species

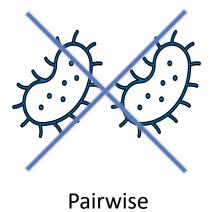


> Conclusion

Tradeoff









Resilience



















Perspective















- David Sherman
- Clémence Frioux
- Simon Labarthe
- Coralie Muller

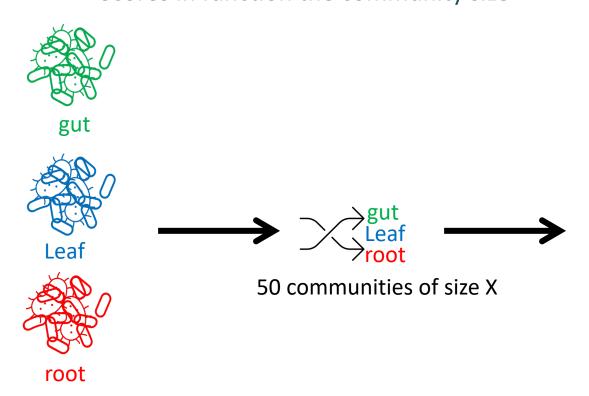
- Hélène Falentin

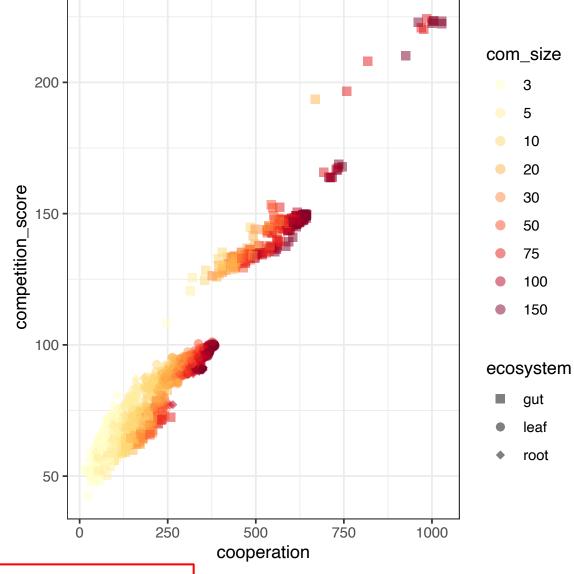
Thanks for your attention



INRAe

Benchmarks for testing scoresScores in function the community size





Reference genomes of cultivable species from 3 ecosystems

Community size is not the main actor in the score variations

Zou, Y., et al, 2019, *Nature Biotechnology*

