— ECML-PKDD'2012 —

Workshop on Learning and Discovery in Symbolic Systems Biology

Abducing Biological Regulatory Networks from Process Hitting models

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Joint work with: Loïc PAULEVÉ³, Katsumi INOUE², Morgan MAGNIN¹, Olivier ROUX¹

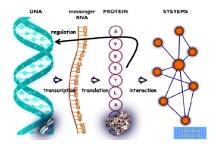
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² Inoue Laboratory / NII / Sokendai University (Tokyo, Japan) ki@nii.ac.jp

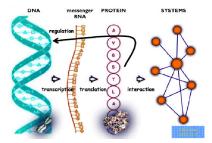
³ AMIB / LIX / École Polytechnique (Palaiseau, France) pauleve@lix.polytechnique.fr

AtlanSTIC sojourn financed by NII & Centrale Initiatives

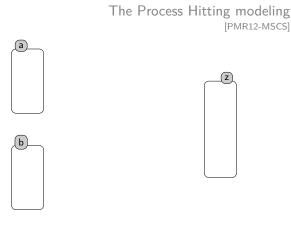
Algebraic modeling to study complex dynamical biological systems:



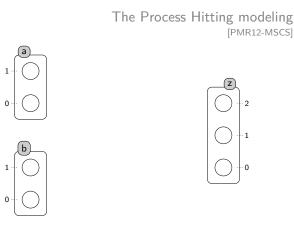
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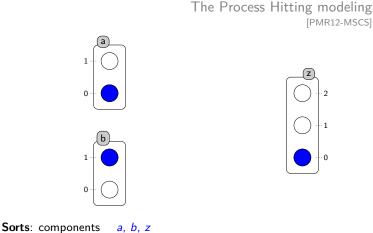
- Historical model: Biological Regulatory Network (René Thomas)
- New developed model: Process Hitting
- \Rightarrow Allow efficient translation from Process Hitting to BRN \mid \Leftarrow



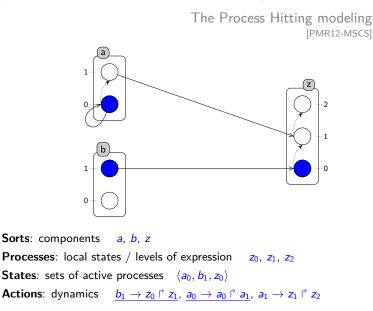
Sorts: components a, b, z

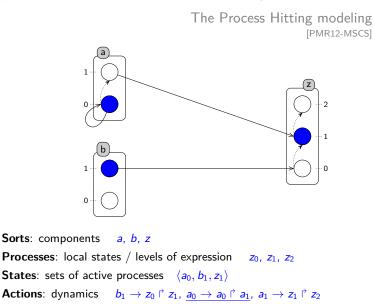


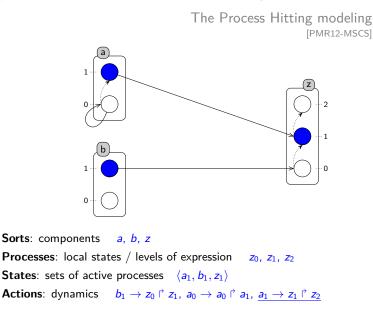
Sorts: components *a*, *b*, *z* **Processes**: local states / levels of expression z_0 , z_1 , z_2

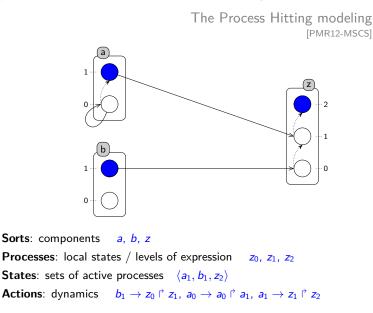


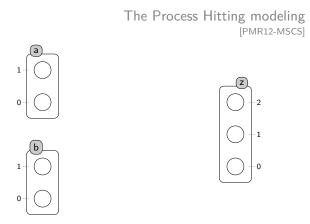
Processes: local states / levels of expression z_0, z_1, z_2 **States:** sets of active processes $\langle a_0, b_1, z_0 \rangle$



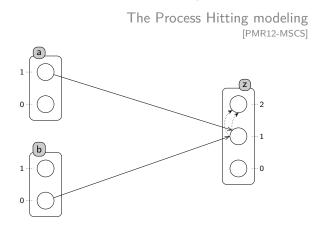




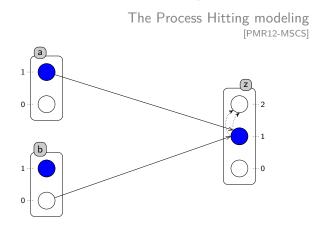




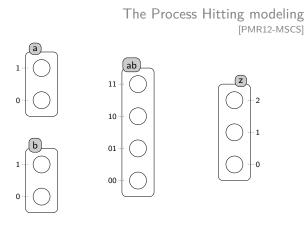
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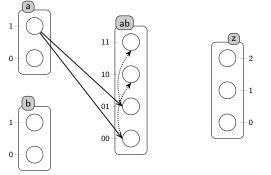
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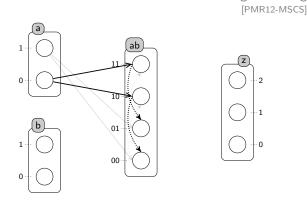


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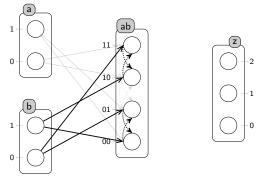


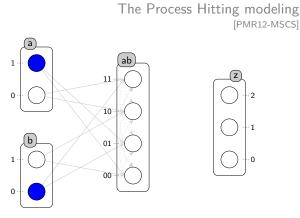




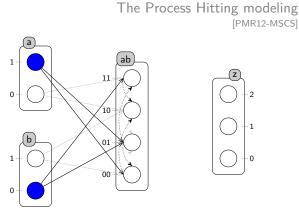
The Process Hitting modeling



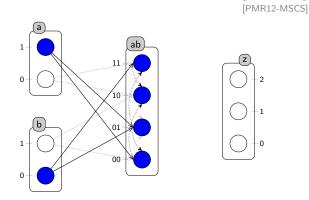




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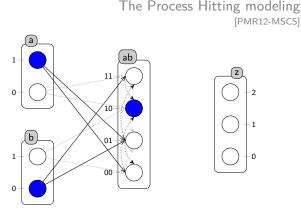


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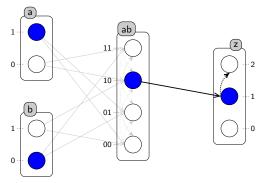
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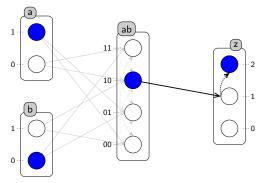
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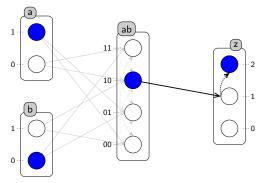
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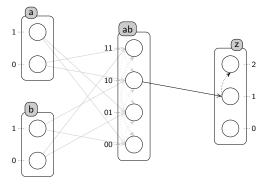
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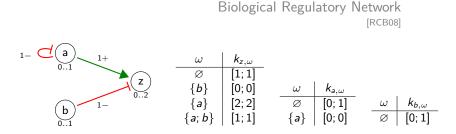
How to introduce some **cooperation** between sorts? $a_1 \wedge b_0 \rightarrow z_1 \uparrow^2 z_2$ Solution: a **cooperative sort** ab to express $a_1 \wedge b_0$ Constraint: each configuration is represented by one process $\langle a_1, b_0 \rangle \Rightarrow ab_{10}$ Advantage: regular sort; drawbacks: complexity, temporal shift



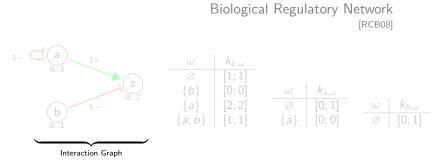


The Process Hitting framework:

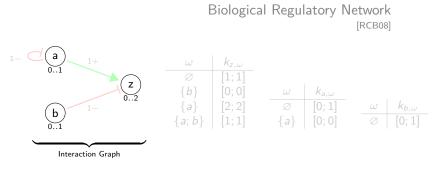
- Dynamic modeling with an atomistic point of view
- Efficient static analysis (fixed points, reachability)
- Possible extensions (stochasticity, priorities)
- Useful for the study of large biological models



Historical bio-informatics model for studying genes interactions Widely used and well-adapted to represent dynamic gene systems



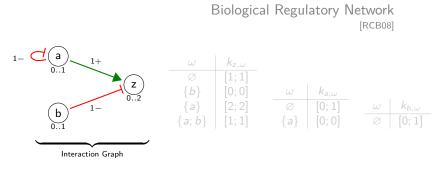
Interaction Graph: structure of the system (genes & interactions)



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Nodes: genes

- \rightarrow Name *a*, *b*, *z*
- \rightarrow Possible values (levels of expression) 0..1, 0..2



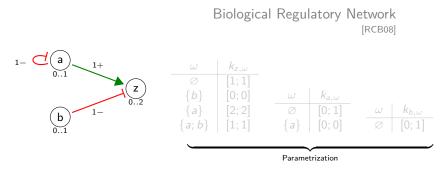
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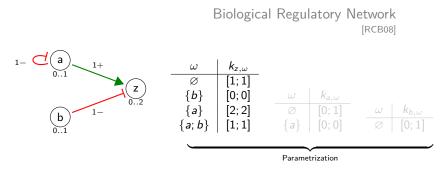
- \rightarrow Name *a*, *b*, *z*
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Edges: interactions

- ightarrow Type (activation or inhibition) ightarrow + / -
- \rightarrow Threshold 1



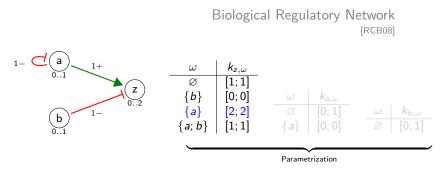
Parametrization: strength of the influences (evolution tendencies)



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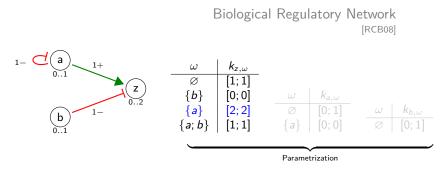
Maps of tendencies for each gene

- ightarrow To any set of predecessors ω
- \rightarrow Corresponds a **parameter** $k_{x,\omega}$



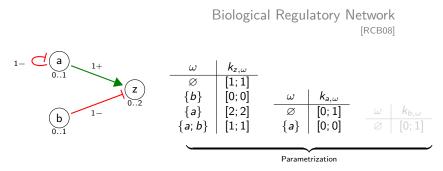
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 $\begin{array}{l} \text{Maps of tendencies for each gene} \\ \rightarrow \text{ To any set of predecessors } \quad \omega \\ \rightarrow \text{ Corresponds a parameter } \quad k_{\mathrm{x},\omega} \\ \quad ``k_{\mathrm{z},\{a\}} = [2;2]'' \quad \text{ means: } \quad ``z \text{ tends to } [2;2] \text{ when } a \geq 1 \text{ and } b < 1'' \end{array}$



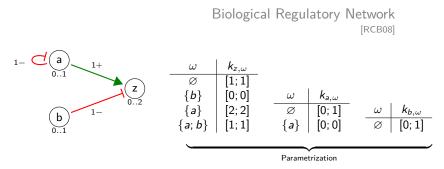
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Maps of tendencies for each gene \rightarrow To any **set of predecessors** ω \rightarrow Corresponds a **parameter** $k_{x,\omega}$ " $k_{z,\{a\}} = [2;2]$ " means: "z tends to 2 when a = 1 and b = 0"



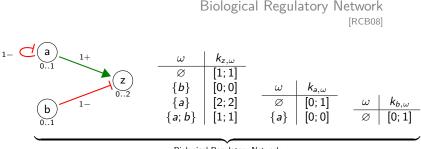
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Biological Regulatory Network

- \rightarrow All needed information to run the model or study its dynamics:
 - Build the State Graph
 - Find reachability properties, fixed points, attractors
 - Other properties...
- ightarrow Strengths: well adapted for the study of biological systems
- → **Drawbacks**: inherent complexity; needs the full specification of cooperations

ASP Implementation

ASP: Declarative programming

Rule: $head \leftarrow body$. Fact: head. Constraint: $\leftarrow body$. Aggregate: $lower \{ atoms \} upper \leftarrow body$. Abducing BRNs from PH models o Frameworks Definitions o Answer Set Programming

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Representation of PH / BRNs:

Gene: component(a, n).

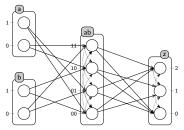
Action: action(a, i, b, j, k).

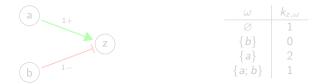
Cooperation: cooperation(c, a, i, j).

Useful rules: component_levels(X, 0..M) \leftarrow component(X, M).
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Abducing BRNs from PH models o Translating a Process Hitting into a BRN

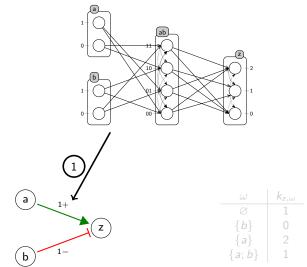
Inferring a BRN with Thomas' parameters [CMSB12]





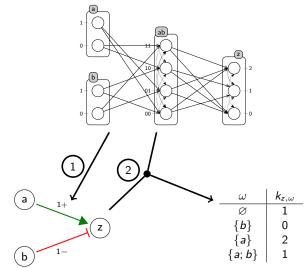
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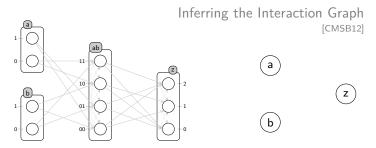
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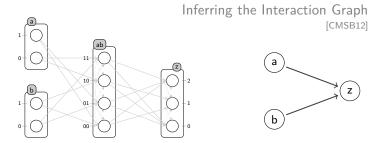
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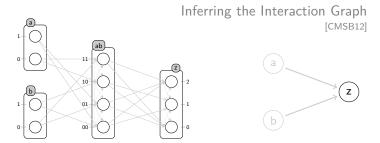
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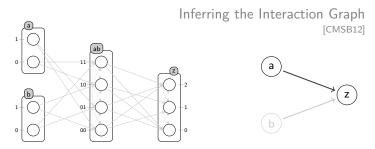


- Inputs: a Process Hitting model
- **Output:** An interaction graph with all information: \rightarrow edges, signs and thresholds
- Difficulties: Process Hitting is more atomistic than BRNs
- Idea: Exhaustive search in all possible configurations

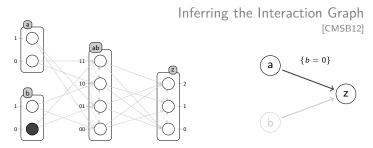




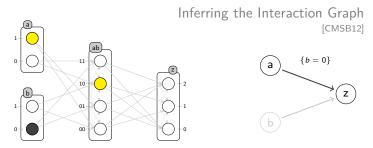
• For each gene [z]



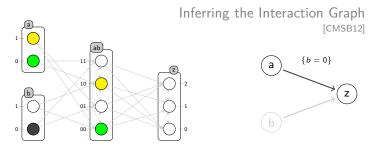
• For each gene [z], consider one possible regulator [a]



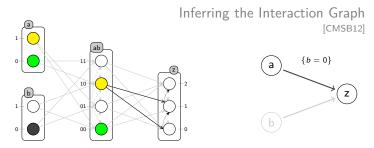
- For each gene [z], consider one possible regulator [a]
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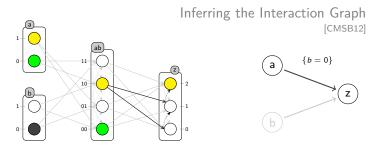
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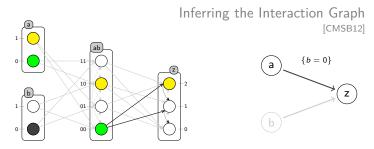
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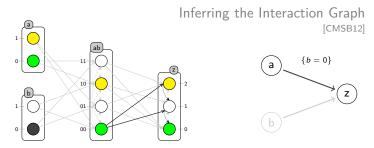
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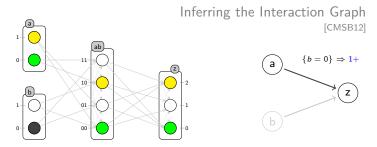
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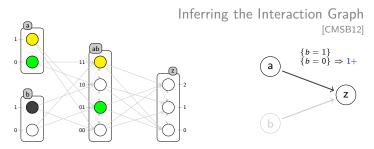
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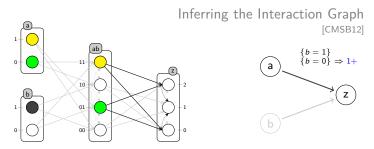
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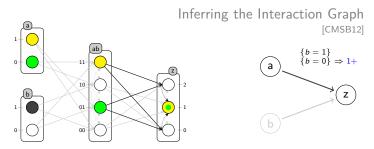
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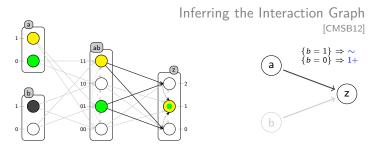
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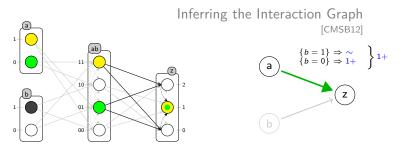


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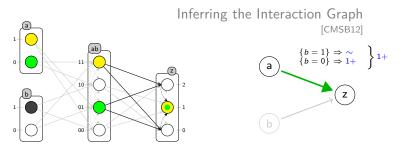
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• If possible, determine the general influence of a on z



• For each gene [z], consider one possible regulator [a]

- Consider a configuration of all other regulators [{b = 1}]
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If possible, determine the general influence of a on z

Problematic cases:

- $\left. \begin{array}{l} \rightarrow \mbox{ No focal processes (cycle)} \\ \rightarrow \mbox{ Opposite influences } (+ \& -) \end{array} \right\} \Rightarrow \mbox{ Unsigned edge}$

Interaction Graph Inference

Implementation

Programming in ASP:

- Formal mathematical definitions \rightarrow ASP
- Use of aggregates (enumeration = 1 active process per sort)

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Calling ASP:

- Pint (existing OCaml library) to read Process Hitting models Free library + examples: http://processhitting.wordpress.com/
- **OCaml** to translate these models to an ASP description and parse the results
- Clingo to solve the description with the adequate program

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Interaction Graph Inference
Results
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Results: Very fast execution (personal laptop, 1.83GHz dual-core)

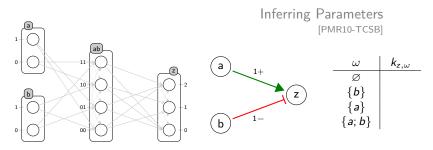
< 1s for 20 & 40 genes models [EGFR20 & TCRSIG40]

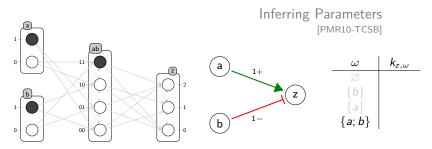
 $\simeq 13s$ for a 94 genes model [TCRSIG94]

 \simeq 4min for a 104 genes model [EGFR104]

Model name	Sorts	Cooperative sorts	Processes	Actions
[EGFR20]	20	22	152	399
[TCRSIG40]	40	14	156	301
[TCRSIG94]	94	39	448	1124
[EGFR104]	104	89	748	2356

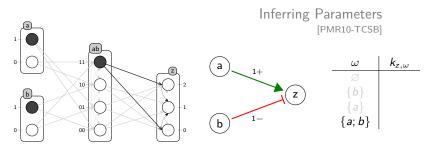
- [EGFR20]: Epidermal Growth Factor Receptor, by Özgür Sahin et al.
- [EGFR104]: Epidermal Growth Factor Receptor, by Regina Samaga et al.
- [TCRSIG40]: T-Cell Receptor Signaling, by Steffen Klamt et al.
- [TCRSIG94]: T-Cell Receptor Signaling, by Julio Saez-Rodriguez et al.



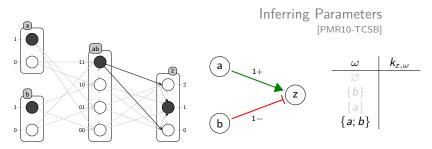


Inputs: The Process Hitting model and the related Interaction Graph **Output:** The Parametrization related to the Interaction Graph

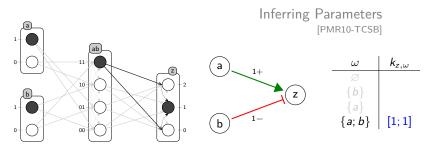
• For each gene [z] and each **configuration** of resources $[\omega = \{a; b\}]$



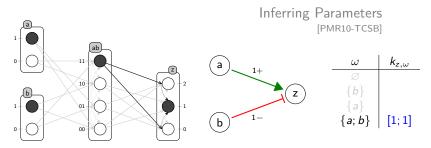
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- Under some conditions, this set is the parameter: $k_{z,\{a,b\}} = [1;1]$

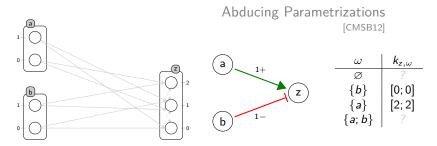


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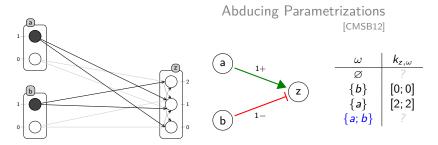
- For each gene [z] and each **configuration** of resources $[\omega = \{a; b\}]$
- Find the set of **focal processes** of the gene [{z₁}]
- Under some <u>conditions</u>, this set is the parameter: $k_{z,\{a,b\}} = [1;1]$

Problematic cases:

- ightarrow Behavior cannot be represented as a BRN
- ightarrow Lack of cooperation (no focal processes)

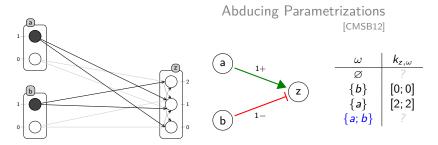


Inputs: The Process Hitting, the related Interaction Graph and the partially inferred ParametrizationOutput: All admissible Parametrizations observing the dynamics



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• Incomplete cooperations may lead to a partial Parametrization $[\omega = \{a, b\}]$

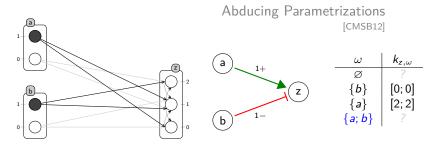


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Abducing BRNs from PH models o Translating a Process Hitting into a BRN o Parametrization Inference



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- \rightarrow Enumeration regarding:
 - Biological constraints
 - The dynamics of the Process Hitting

Abducing BRNs from PH models o Translating a Process Hitting into a BRN o Parametrization Inference

Abducing Parametrizations

Implementation

Parameters definitions:

One identifier for each parameter: param_label(a, i)

Useful rules:

less_active(X, P, Q) $\leftarrow K_{X,P}$ has less activators than $K_{X,Q}$ *param_inf*(X, P, Q) $\leftarrow K_{X,P} \preccurlyeq K_{X,Q}$

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Parametrizations filtering uses constraints: $\leftarrow less_active(X, P, Q), \neg param_inf(X, P, Q).$ [X: component; P, Q: parameter labels]

Maxime FOLSCHETTE

Abducing BRNs from PH models o Translating a Process Hitting into a BRN o Parametrization Inference

Parametrization Inference Results

Two steps:

- Parameters inference (partial)
- Parametrization abduction (total)

Abducing BRNs from PH models \circ Translating a Process Hitting into a BRN \circ Parametrization Inference

Parametrization Inference

Two steps:

- Parameters inference (partial)
- Parametrization abduction (total)

Results:

• Very fast execution for parameters inference

< 1s for 20 & 40 genes models [EGFR20 & TCRSIG40]

Parametrization abduction

After one cooperation removal:

 \simeq 4s to find 42 admissible Parametrizations [TCRSIG40]

 \simeq 20s to find 129 admissible Parametrizations [EGFR20]

ASP is convenient to handle enumeration (cardinalities) and filter only admissible answers (constraints)

Summary & Future work

- Inference of the complete Interaction Graph
 - \rightarrow Exhaustive approach to find the mutual influences
- Inference of the possibly partial Parametrization
 - \rightarrow Exhaustive approach to find the necessary parameters
- Abduce all full & admissible Parametrizations
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- Abduce all full & admissible Parametrizations
 - \rightarrow Exhaustive approach to find only relevant answers
- Complexity: linear in the number of genes, exponential in the number of regulators of one gene
- Concretize into more expressive BRN representations
 - \rightarrow Tackle with **unsigned edges** (problematic cases)
 - \rightarrow Use multiplexes to decrease the size of Parametrizations
- Use projections to remove cooperative sorts
 - \rightarrow Make actions independent
 - \rightarrow Drop inference complexity?

Conclusion

Existing translation: René Thomas → Process Hitting New translation: Process Hitting → René Thomas

- \rightarrow New formal link between the two models
- \rightarrow More **visibility** to the Process Hitting

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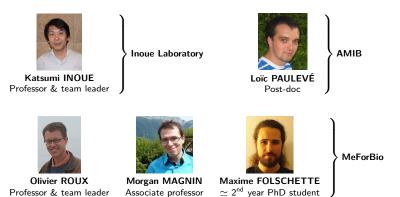
- \rightarrow New formal link between the two models
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Using ASP

- \rightarrow Tackles with complexity/combinatorial explosion
- \rightarrow Allows efficient exhaustive search & enumeration

A multi-team topic

Inoue Laboratory (NII, Sokendai): Constraint Programming, Systems Biology **MeForBio** (IRCCyN, ÉCN): Formal Methods for Bioinformatics **AMIB** (LIX, Polytechnique): Algorithms and Models for Integrative Biology



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Thank you