

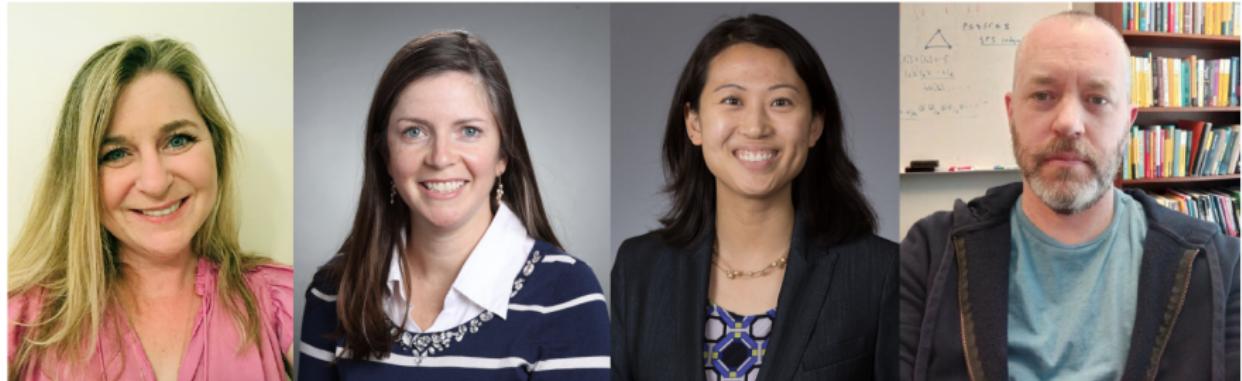
Identifiability of Linear Compartmental Tree Models

Cash Bortner

CSU-Stanislaus
cbortner@csustan.edu

DART XI
Queen Mary University of London
6/7/2023 or 7/6/2023

Collaborators



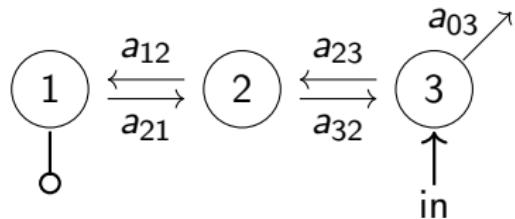
Dr. Elizabeth Gross
U. of Hawai'i at Mānoa

Dr. Nicolette Meshkat
Santa Clara University

Dr. Anne Shiu
Texas A&M University

Dr. Seth Sullivant
North Carolina State U.

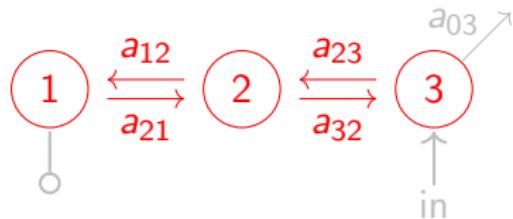
Motivating Example



Linear Compartmental Model

$$\begin{aligned}\mathcal{M} &= (G, In, Out, Leak) \\ &= (\text{Cat}_3, \{3\}, \{1\}, \{3\}).\end{aligned}$$

Motivating Example

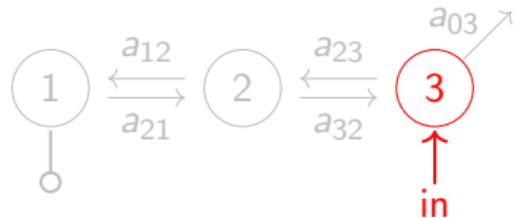


Linear Compartmental Model

$$\begin{aligned}\mathcal{M} &= (\textcolor{red}{G}, \textcolor{gray}{In}, \textcolor{gray}{Out}, \textcolor{gray}{Leak}) \\ &= (\textcolor{red}{Cat}_3, \{3\}, \{1\}, \{3\}).\end{aligned}$$

Directed Graph: $G = \text{Cat}_3$

Motivating Example

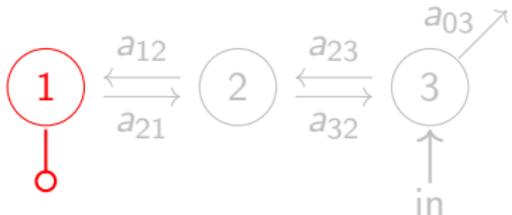


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Input Compartment: $In = \{3\}$

Motivating Example

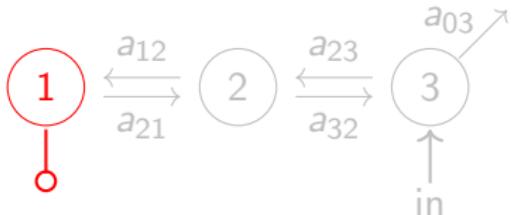


Linear Compartmental Model

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Measured Compartment: $Out = \{1\}$

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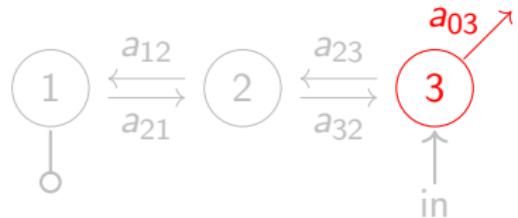


Linear Compartmental Model

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“Output” Compartment: $Out = \{1\}$

Motivating Example

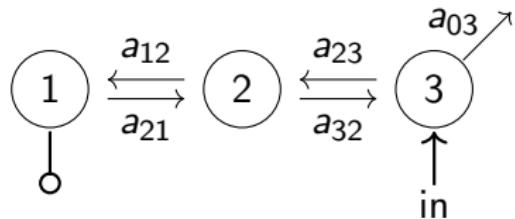


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Leak Compartment: $\text{Leak} = \{3\}$

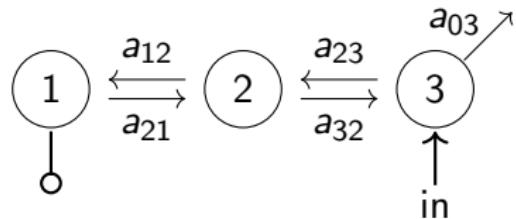
Motivating Example



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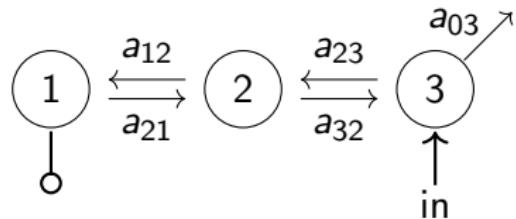
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Motivating Question: Identifiability

Given information about the input and output compartment[s], can we **recover** all flow rate parameters?

Motivating Example



Linear Compartmental Model

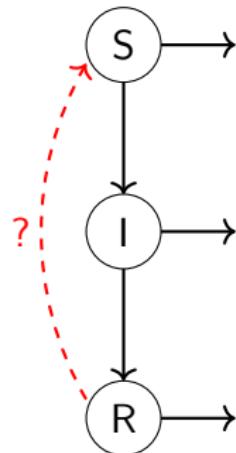
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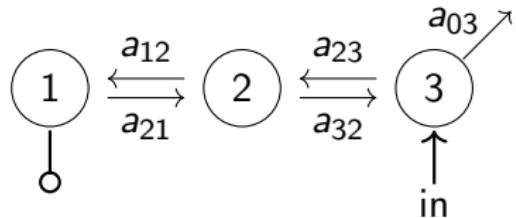
Given information about the input and output compartment[s], can we **identify** all flow rate parameters?

Compartmental Models in the Wild

- SIR Model for spread of a virus in Epidemiology (non-linear)
- SIV Model for vaccine efficiency in Epidemiology
- Modeling Pharmacokinetics for absorption, distribution, metabolism, and excretion in the blood
- Modeling different biological systems

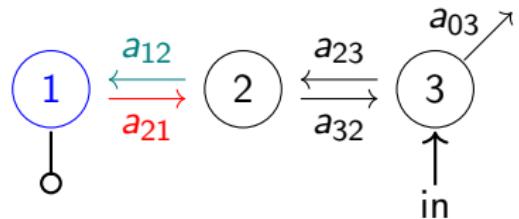


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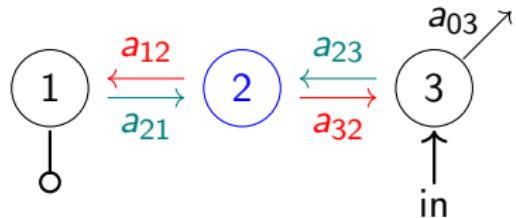


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ODEs in terms of concentrations $x_i(t)$, input $u_3(t)$, and output $y_1(t)$:

$$\dot{x}_1 = -a_{21}x_1(t) + a_{12}x_2(t)$$

Motivating Example



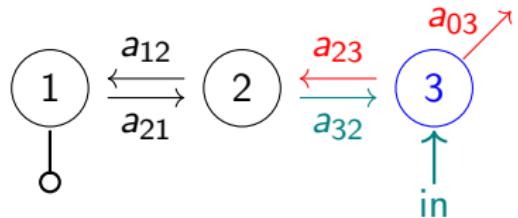
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$$\dot{x}_2 = a_{21}x_1(t) - (a_{12} + a_{32})x_2(t) + a_{23}x_3(t)$$

Motivating Example



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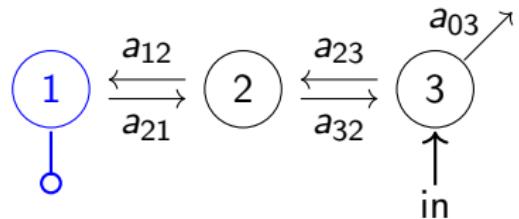
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Motivating Example



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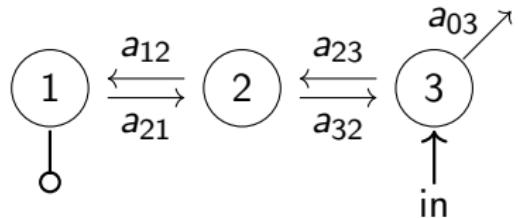
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$$y_1(t) = x_1(t).$$

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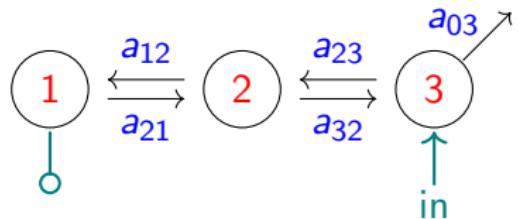
ODEs in terms of concentrations $x_i(t)$, input $u_3(t)$, and output $y_1(t)$:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{pmatrix} = \underbrace{\begin{pmatrix} -a_{21} & a_{12} & 0 \\ a_{21} & -a_{12} - a_{32} & a_{23} \\ 0 & a_{32} & -a_{03} - a_{23} \end{pmatrix}}_{\text{compartmental matrix A}} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ u_3(t) \end{pmatrix}$$

with

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Motivating Example



$$\begin{aligned}\mathcal{M} &= (G, In, Out, Leak) \\ &= (\text{Cat}_3, \{3\}, \{1\}, \{3\}).\end{aligned}$$

Goal: Identify the parameters a_{ji} from the measurable variables.

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Motivating Example

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$$\begin{pmatrix} \partial_t & 0 & 0 \\ 0 & \partial_t & 0 \\ 0 & 0 & \partial_t \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} = \begin{pmatrix} -a_{21} & a_{12} & 0 \\ a_{21} & -a_{12} - a_{32} & a_{23} \\ 0 & a_{32} & -a_{03} - a_{23} \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ u_3(t) \end{pmatrix}$$

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Motivating Example

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$$\begin{pmatrix} \partial_t + a_{21} & -a_{12} & 0 \\ -a_{21} & \partial_t + a_{12} + a_{32} & -a_{23} \\ 0 & -a_{32} & \partial_t + a_{03} + a_{23} \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ u_3(t) \end{pmatrix}$$

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via Cramer's Rule:

$$\begin{aligned} \det \begin{pmatrix} \partial_t + a_{21} & -a_{12} & 0 \\ -a_{21} & \partial_t + a_{12} + a_{32} & -a_{23} \\ 0 & -a_{32} & \partial_t + a_{03} + a_{23} \end{pmatrix} y_1(t) \\ = \det \begin{pmatrix} 0 & -a_{12} & 0 \\ 0 & \partial_t + a_{12} + a_{32} & -a_{23} \\ u_3(t) & -a_{32} & \partial_t + a_{03} + a_{23} \end{pmatrix} \end{aligned}$$

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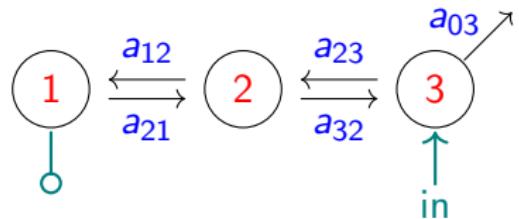
Goal: Identify the parameters a_{ji} from the measurable variables.

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Motivating Example



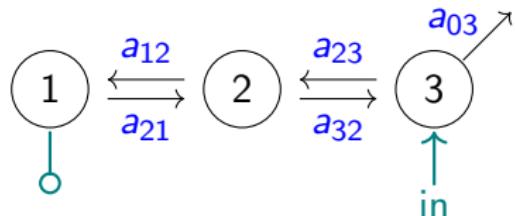
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Goal: Identify the *parameters* a_{ji} from the *measurable variables*.

$$\det(\partial_t I - A)y_1 = \underbrace{\det(\partial_t I - A)^{(3,1)}}_{\text{remove row 3 and col 1}} u_3$$

by Cramer's Rule and substitution.

Motivating Example: Input/Output Equation



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Via a substitution and application of Cramer's Rule:

$$\begin{aligned}y_1^{(3)} + (a_{03} + a_{12} + a_{21} + a_{23} + a_{32})\ddot{y}_1 + (a_{03}a_{12} + a_{03}a_{21} \\ + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32})\dot{y}_1 + (a_{03}a_{21}a_{32})y_1 = (a_{12}a_{23})u_3.\end{aligned}$$

an ODE in only the **measurable variables** and the **parameters**:

Input/Output Equation

Goal: Identify **parameters a_{ji}** from the **measurable variables**.

Identifiability Analysis: Structural vs. Practical

Overview

We want to recover (identify) parameters of ODE models from measured variables.

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Remark

Structural identifiability is a **necessary condition** for practical identifiability.

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Structural identifiability is a **necessary condition** for practical identifiability.

Structural Identifiability Analysis: A two part problem

Structural Identifiability via the input-output equation

We consider structural identifiability as a two-step problem:

1. Find an input/output equation of the ODE system in terms of measurable variables
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Goal

We want to classify structural identifiability

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We consider structural identifiability as a two-step problem:

1. Find an input/output equation of the ODE system in terms of measurable variables
2. Determine the injectivity of the coefficient map defined by the input/output equation

Goal

*We want to classify structural identifiability by the **underlying graph structure.***

Novel Input-Output Equation Characterization

Theorem (\$, Gross, Meshkat, Shiu, Sullivant [1])

The coefficients of the input-output equation of a LCM ($G, In, Out, Leak$) can be generated by *incoming forests* on graphs related to G .

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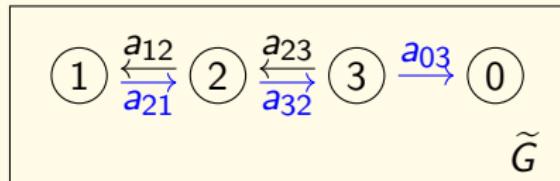
Definitions

A directed graph H is called an *incoming forest* if

- no vertex has more than one outgoing edge, and
- its underlying undirected graph is a forest

Example

The set of incoming forests with 3 edges on \tilde{G} : $\mathcal{F}_3(\tilde{G}) = \{\{1 \rightarrow 2, 2 \rightarrow 3, 3 \rightarrow 0\}\}$



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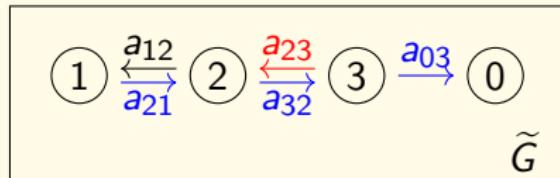
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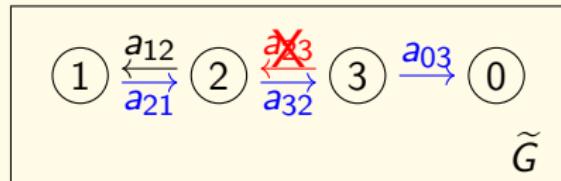
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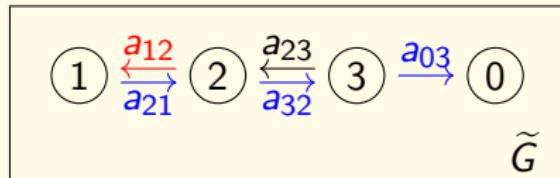
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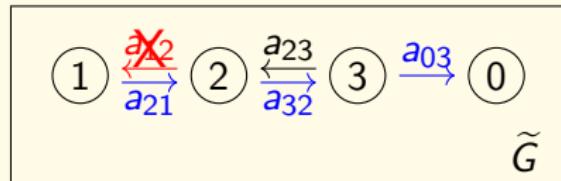
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- no vertex has more than one outgoing edge, and
- its underlying undirected graph is a forest

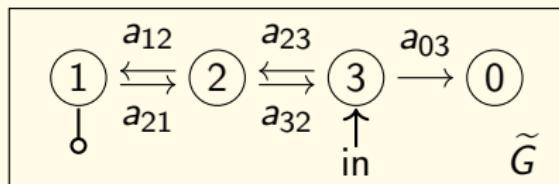
Example

The set of incoming forests with 3 edges on \tilde{G} : $\mathcal{F}_3(\tilde{G}) = \{\{1 \rightarrow 2, 2 \rightarrow 3, 3 \rightarrow 0\}\}$



Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$:



The k^{th} coefficient of LHS of the i-o equation is:

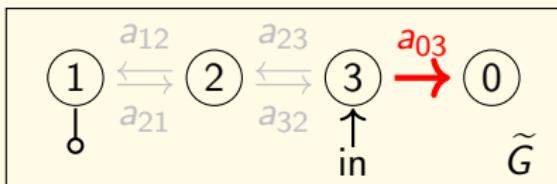
$$c_k = \sum_{F \in \mathcal{F}_{3-k}(\tilde{G})} \pi_F$$

LHS coefficients: Incoming forests with 1 edge

Derivative	Coefficient
$y_1^{(3)}$	1
$y_1^{(2)}$	$a_{03} + a_{12} + a_{21} + a_{23} + a_{32}$
$y_1^{(1)}$	$a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32}$
$y_1^{(0)}$	$a_{03}a_{21}a_{32}$

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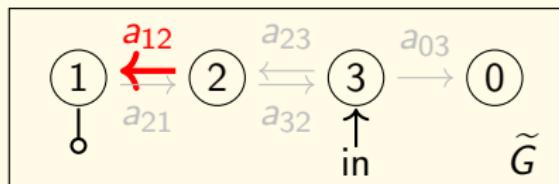
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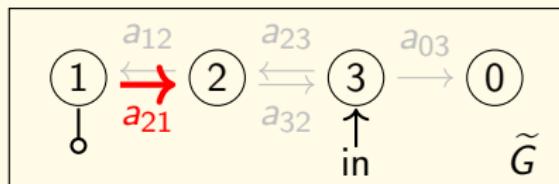
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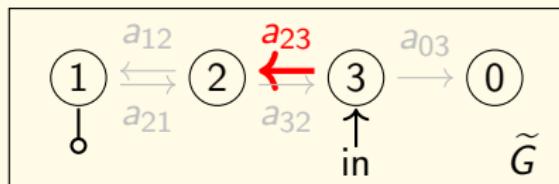
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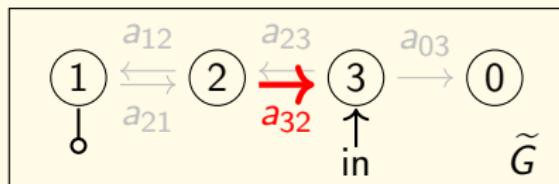
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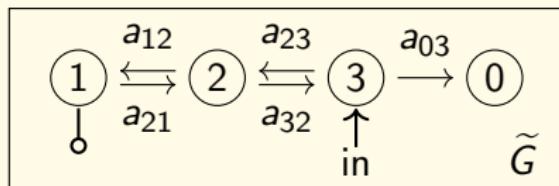
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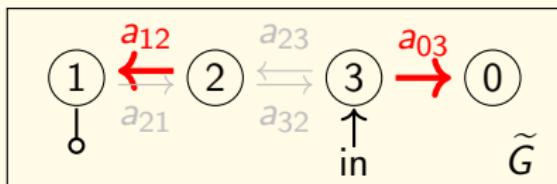
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LHS coefficients: Incoming forests with 2 edge

Derivative	Coefficient
$y_1^{(3)}$	1
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$y_1^{(1)}$	$a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32}$
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Example

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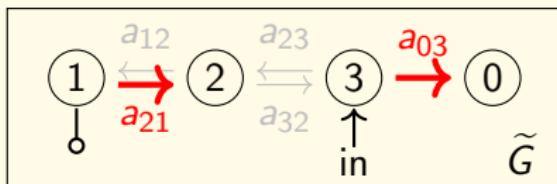
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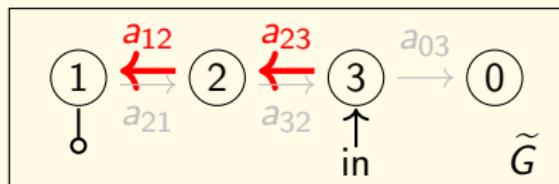
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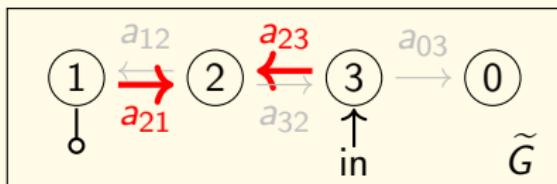
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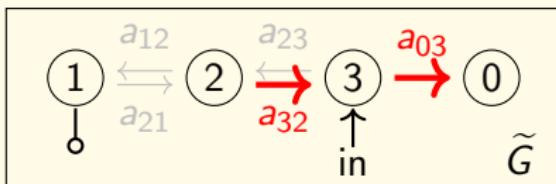
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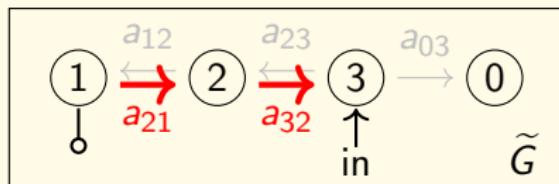
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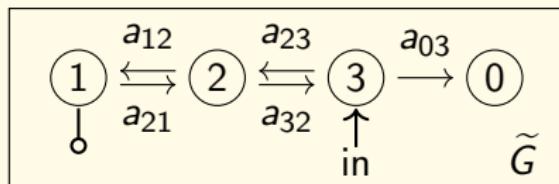
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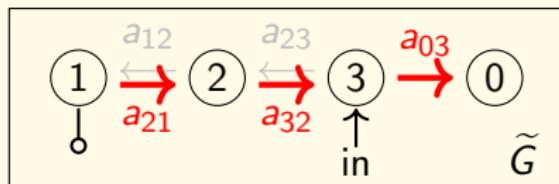
$$c_k = \sum_{F \in \mathcal{F}_{3-k}(\tilde{G})} \pi_F$$

LHS coefficients: Incoming forests with 3 edge

Derivative	Coefficient
$y_1^{(3)}$	1
$y_1^{(2)}$	$a_{03} + a_{12} + a_{21} + a_{23} + a_{32}$
$y_1^{(1)}$	$a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32}$
$y_1^{(0)}$	$a_{03}a_{21}a_{32}$

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$:



The k^{th} coefficient of LHS of the i-o equation is:

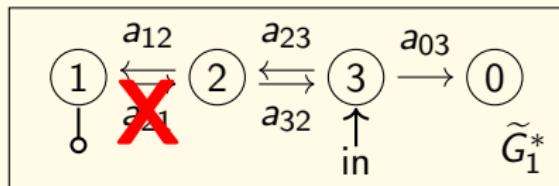
$$c_k = \sum_{F \in \mathcal{F}_{3-k}(\tilde{G})} \pi_F$$

LHS coefficients: Incoming forests with 3 edge

Derivative	Coefficient
$y_1^{(3)}$	1
$y_1^{(2)}$	$a_{03} + a_{12} + a_{21} + a_{23} + a_{32}$
$y_1^{(1)}$	$a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32}$
$y_1^{(0)}$	$a_{03}a_{21}a_{32}$

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$:



The k^{th} coefficient of RHS of the i-o equation is:

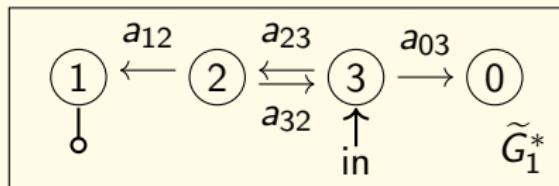
$$d_k = \sum_{F \in \mathcal{F}_{3-k-1}^{3,1}(\tilde{G}_1^*)} \pi_F$$

RHS coefficients:

Derivative	Coefficient
$u_3^{(0)}$	

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$:



The k^{th} coefficient of RHS of the i-o equation is:

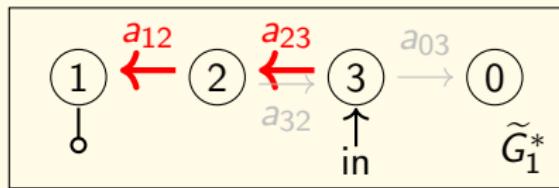
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RHS coefficients: Incoming forests with 2 edges AND a path from 3 to 1

Derivative	Coefficient
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For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$:



The k^{th} coefficient of RHS of the i-o equation is:

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RHS coefficients: Incoming forests with 2 edges AND a path from 3 to 1

Derivative	Coefficient
$u_3^{(0)}$	$a_{12}a_{23}$

Number of Coefficients

Corollary (\$, Gross, Meshkat, Shiu, Sullivant [1])

Consider $\mathcal{M} = (G, \{in\}, \{out\}, Leak)$ where G is strongly connected and $|V_G| = n$. Then the number of non-trivial coefficients in the input/output equation is:

$$\# \text{ on LHS} = \begin{cases} n & \text{if } |Leak| \neq 0 \\ n - 1 & \text{if } |Leak| = 0 \end{cases}, \quad \# \text{ on RHS} = \begin{cases} n - 1 & \text{if } in = out \\ n - \text{dist}(in, out) & \text{if } in \neq out. \end{cases}$$

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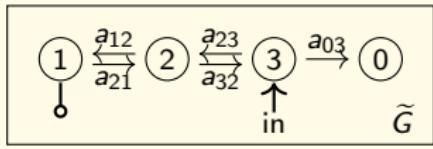
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Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$, the input/output equation is:

$$y_1^{(3)} + (a_{03} + a_{12} + a_{21} + a_{23} + a_{32})y_1'' + (a_{03}a_{12} + a_{03}a_{21} \\ a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32})y_1' + (a_{03}a_{21}a_{32})y_1 = (a_{12}a_{23})u_3.$$



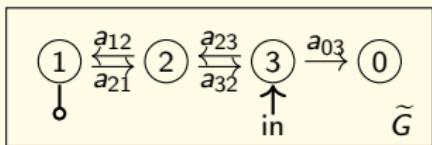
$$\# \text{ on LHS} = 3 \text{ (since } |Leak| = 1\text{)} \\ \# \text{ on RHS} = 3 - \underbrace{\text{dist}(3, 1)}_2 = 1$$

Example

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For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$, the input/output equation is:

$$y_1^{(3)} + (a_{03} + a_{12} + a_{21} + a_{23} + a_{32})\ddot{y}_1 + (a_{03}a_{12} + a_{03}a_{21} \\ + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32})\dot{y}_1 + (a_{03}a_{21}a_{32})y_1 = (a_{12}a_{23})u_3.$$



on LHS = 3

on RHS = 1

The *coefficient map* corresponding to \mathcal{M} is:

$$\phi_{\mathcal{M}}: \mathbb{R}^5 \rightarrow \mathbb{R}^4$$

$$\begin{pmatrix} a_{03} \\ a_{12} \\ a_{21} \\ a_{23} \\ a_{32} \end{pmatrix} \mapsto \begin{pmatrix} a_{03} + a_{12} + a_{21} + a_{23} + a_{32} \\ a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32} \\ a_{03}a_{21}a_{32} \\ a_{12}a_{23} \end{pmatrix}$$

Identifiability

Definition*

A model $(G, In, Out, Leak)$ with coefficient map ϕ is

- *locally identifiable* (identifiable) if, outside a set of measure zero, every point in $\mathbb{R}^{|E_G|+|Leak|}$ has an open neighborhood U for which the restriction $\phi|_U : U \rightarrow \mathbb{R}^m$ is **one-to-one**; and
- *unidentifiable* if c is generically **infinite-to-one**.

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Proposition (Sufficient condition for unidentifiability)

A model $\mathcal{M} = (G, In, Out, Leak)$ is *unidentifiable* if

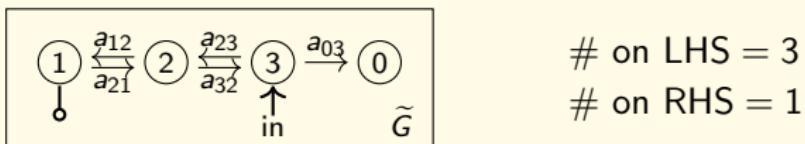
$$\underbrace{\# \text{ parameters}}_{|E_G|+|Leak|} > \# \text{ coefficients.}$$

Example

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$, the input/output equation is:

$$y_1^{(3)} + (a_{03} + a_{12} + a_{21} + a_{23} + a_{32})\ddot{y}_1 + (a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32})\dot{y}_1 + (a_{03}a_{21}a_{32})y_1 = (a_{12}a_{23})u_3.$$



The coefficient map corresponding to \mathcal{M} is:

$$\phi_{\mathcal{M}}: \mathbb{R}^5 \rightarrow \mathbb{R}^4$$

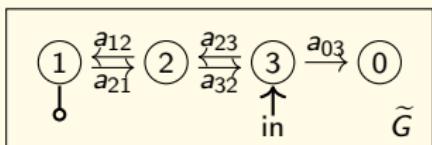
$$\begin{pmatrix} a_{03} \\ a_{12} \\ a_{21} \\ a_{23} \\ a_{32} \end{pmatrix} \mapsto \begin{pmatrix} a_{03} + a_{12} + a_{21} + a_{23} + a_{32} \\ a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32} \\ a_{03}a_{21}a_{32} \\ a_{12}a_{23} \end{pmatrix}$$

Example

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$, the input/output equation is:

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on LHS = 3

on RHS = 1

The coefficient map corresponding to \mathcal{M} is:

$$\phi_{\mathcal{M}}: \mathbb{R}^5 \rightarrow \mathbb{R}^4 \quad \mathcal{M} \text{ is UNIDENTIFIABLE}$$

$$\begin{pmatrix} a_{03} \\ a_{12} \\ a_{21} \\ a_{23} \\ a_{32} \end{pmatrix} \mapsto \begin{pmatrix} a_{03} + a_{12} + a_{21} + a_{23} + a_{32} \\ a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32} \\ a_{03}a_{21}a_{32} \\ a_{12}a_{23} \end{pmatrix}$$

Unidentifiability

Corollary (\$, Gross, Meshkat, Shiu, Sullivant [1])

Consider $\mathcal{M} = (G, \{in\}, \{out\}, \text{Leak})$ where G is strongly connected and $|V_G| = n$. Define L and d as follows:

$$L = \begin{cases} 0 & \text{if } |\text{Leak}| = 0 \\ 1 & \text{if } |\text{Leak}| \neq 0 \end{cases} \quad \text{and} \quad d = \begin{cases} 1 & \text{if } \text{dist}(in, out) = 0 \\ \text{dist}(in, out) & \text{if } \text{dist}(in, out) \neq 0. \end{cases}$$

Then \mathcal{M} is **unidentifiable** if

$$\underbrace{|\text{Leak}| + |E_G|}_{\# \text{ parameters}} > \underbrace{2n - L - d}_{\# \text{ coefficients}}.$$

The Jacobian

Proposition

$\mathcal{M} = (G, \{i\}, \{j\}, \text{Leak})$ is locally identifiable if and only if the rank of the Jacobian matrix of its coefficient map is equal to # parameters.

Example

For $\mathcal{M} = (\text{Cat}_3, \{3\}, \{1\}, \{3\})$, the input/output equation is:

$$y_1^{(3)} + \underbrace{(a_{03} + a_{12} + a_{21} + a_{23} + a_{32})}_{c_2} y_1 + \underbrace{(a_{03}a_{12} + a_{03}a_{21} + a_{12}a_{23} + a_{21}a_{23} + a_{03}a_{32} + a_{21}a_{32})}_{c_1} y_1 \\ + \underbrace{(a_{03}a_{21}a_{32})}_{c_0} y_1 = \underbrace{(a_{12}a_{23})}_{d_0} u_3$$

$$J(\phi_{\mathcal{M}}) = \begin{pmatrix} c_2 & a_{03} & a_{12} & a_{21} & a_{23} & a_{32} \\ c_1 & a_{12} + a_{21} + a_{32} & a_{03} + a_{23} & a_{03} + a_{23} + a_{32} & a_{12} + a_{21} & a_{03} + a_{21} \\ c_0 & a_{21}a_{32} & 0 & a_{03}a_{32} & 0 & a_{03}a_{21} \\ d_0 & 0 & a_{23} & 0 & a_{12} & 0 \end{pmatrix}$$

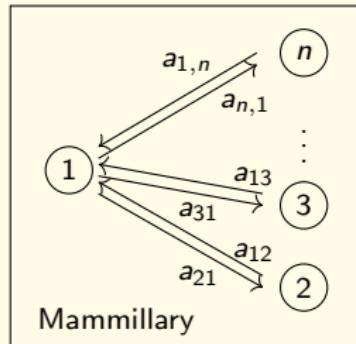
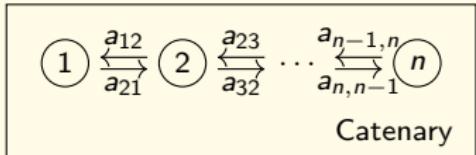
Tree Models

Definition

A (bidirectional) *tree model* $\mathcal{M} = (G, In, Out, \text{Leak})$ has properties

- the edge $i \rightarrow j \in E_G$ if and only if the edge $j \rightarrow i \in E_G$
- underlying undirected graph of G a [double] tree*

Examples



Unidentifiability of Tree Models

Theorem (\$, Gross, Meshkat, Shiu, Sullivant [1])

A tree model $\mathcal{M} = (G, \{in\}, \{out\}, \text{Leak})$ is **unidentifiable** if

$$\text{dist}(\text{in}, \text{out}) \geq 2 \text{ or } |\text{Leak}| \geq 2.$$

Proof idea: Let $n = |V_G|$.

- # parameters: $|E_G| + |\text{Leak}| = 2n - 2 + |\text{Leak}|$

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$\text{dist}(\text{in}, \text{out}) \geq 2$	$2n - \text{dist}(\text{in}, \text{out})$	$2n - \text{dist}(\text{in}, \text{out})$	$2n - \text{dist}(\text{in}, \text{out}) - 1$
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Top-Left:

- # parameters $\geq 2n$ (since $|\text{Leak}| \geq 2$)
- # coefficients $= 2n - \underbrace{\text{dist}(\text{in}, \text{out})}_{\geq 2}$

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Top-Left: UNIDENTIFIABLE

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Bottom-Right:

- # parameters = $2n - 2$ (since $|\text{Leak}| = 0$)
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- five red cases have # parameters > # coefficients \implies unidentifiability

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- **five red cases** have # parameters > # coefficients \implies **unidentifiability**
- **four blue cases** have # parameters = # coefficients,

but that does not guarantee identifiability.

Building Identifiable Tree Models

Plan for showing that # parameters = # coefficients implies identifiability:

- start with some base model that we know is identifiable (Prop*)
- from base model, *build* all tree models where $|Leak| \leq 1$ and $\text{dist}(\text{in}, \text{out}) \leq 1$ and retain identifiability at each step

Proposition* (\$, Gross, Meshkat, Shiu, Sullivant [1])

The tree model $\mathcal{M} = (G, \{i\}, \{i\}, \emptyset)$ is identifiable.

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Proposition* (\$, Gross, Meshkat, Shiu, Sullivant [1])

The tree model $\mathcal{M} = (G, \{i\}, \{i\}, \emptyset)$ is identifiable.

Proposition (Gross, Harrington, Meshkat, Shiu [2])

Let $\mathcal{M} = (G, In, Out, \emptyset)$ be strongly connected and identifiable. Then, the model $\mathcal{M}' = (G, In, Out, \{k\})$ is also identifiable.

Moving the Input/Output

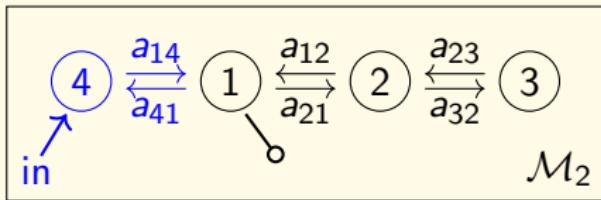
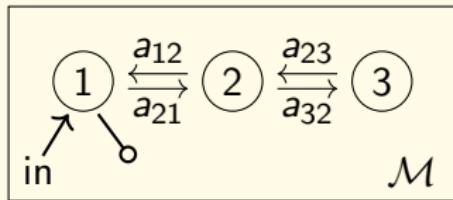
Proposition (\$, Gross, Meshkat, Shiu, Sullivant [1])

Let $\mathcal{M} = (G, \{i\}, \{i\}, \emptyset)$ be an identifiable tree model. Let H be the graph G with the added node n and edges $i \rightarrow n$ and $n \rightarrow i$. Then following models are also identifiable:

- $\mathcal{M}_1 = (H, \{i\}, \{n\}, \emptyset)$
- $\mathcal{M}_2 = (H, \{n\}, \{i\}, \emptyset)$.

Example

Here, $\mathcal{M} = (\text{Cat}_3, \{1\}, \{1\}, \emptyset)$ and $\mathcal{M}_2 = (\text{Cat}_4^*, \{4\}, \{1\}, \emptyset)$:



Proof of Moving the Input/Output

Theorem (\$, Gross, Meshkat, Shiu, Sullivant [1])

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- $\mathcal{M}_1 = (H, \{i\}, \{n\}, \emptyset)$
- $\mathcal{M}_2 = (H, \{n\}, \{i\}, \emptyset)$.

Proof idea:

- write the coefficients of \mathcal{M}_1 in terms of \mathcal{M} and the new parameters
- manipulate the Jacobian of $\phi_{\mathcal{M}_1}$ to “find” the Jacobian of $\phi_{\mathcal{M}}$, which by assumption has full rank:

$$J(\phi_{\mathcal{M}_1}) = \begin{pmatrix} J(\phi_{\mathcal{M}}) & 0 \\ * & C \end{pmatrix}$$

- show that C has full rank using properties of the graph

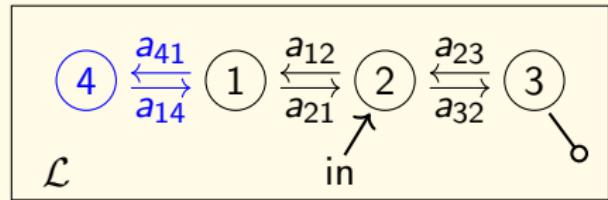
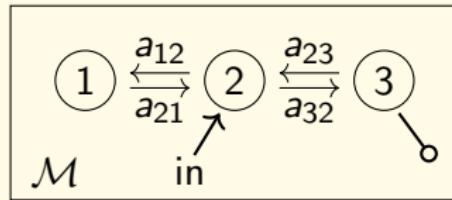
Adding a Leaf

Proposition (\$, Gross, Meshkat, Shiu, Sullivant [1])

Let $\mathcal{M} = (G, \{i\}, \{j\}, \emptyset)$ be an identifiable tree model. Define $\mathcal{L} = (H, \{i\}, \{j\}, \emptyset)$ where H is the graph G with the added node n and edges $k \rightarrow n$ and $n \rightarrow k$ for some $k \in V_G$. Then, \mathcal{L} is identifiable.

Example

Here, $\mathcal{M} = (\text{Cat}_3, \{2\}, \{3\}, \emptyset)$ and $\mathcal{L} = (\text{Cat}_4^*, \{2\}, \{3\}, \emptyset)$:



Classification of Tree Models

Theorem (\$, Gross, Meshkat, Shiu, Sullivant [1])

A tree model $\mathcal{M} = (G, \{in\}, \{out\}, \text{Leak})$ is identifiable if and only if $\text{dist}(in, out) \leq 1$ and $|\text{Leak}| \leq 1$.

Proof outline:

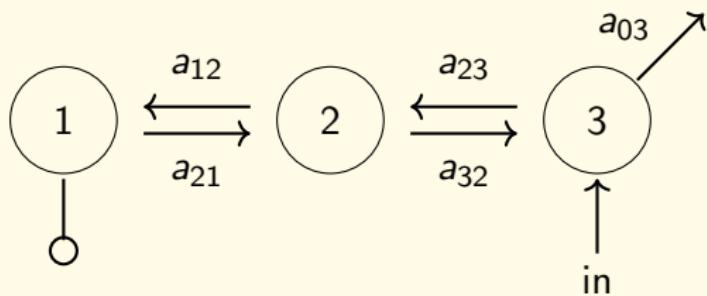
- \mathcal{M} is unidentifiable if either $\text{dist}(in, out) > 1$ or $|\text{Leak}| > 1$
- \mathcal{M} is identifiable if $in = out$ and $|\text{Leak}| = 0$
- \mathcal{M} is identifiable if $\text{dist}(in, out) = 1$ and $|\text{Leak}| = 0$
- if \mathcal{M} is identifiable with $|\text{Leak}| = 0$, then it is identifiable with $|\text{Leak}| = 1$

Example

Theorem (\$, Gross, Meshkat, Shiu, Sullivant [1])

A tree model $\mathcal{M} = (G, \{\text{in}\}, \{\text{out}\}, \text{Leak})$ is identifiable if and only if $\text{dist}(\text{in}, \text{out}) \leq 1$ and $|\text{Leak}| \leq 1$.

Example

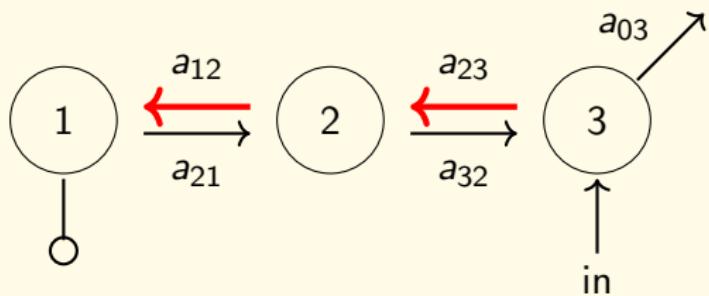


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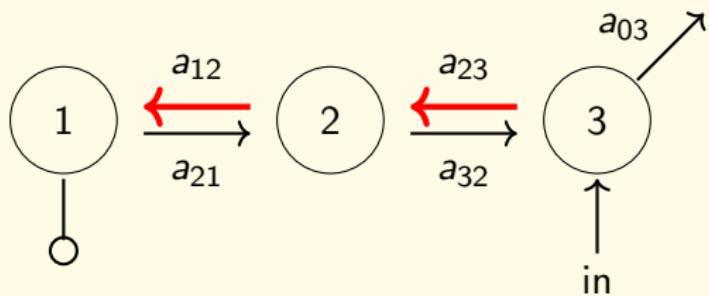
UNIDENTIFIABLE,

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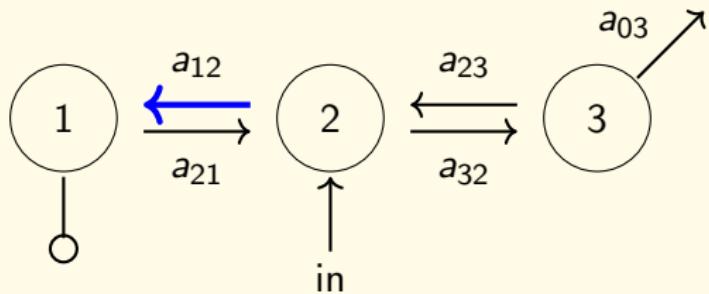
UNIDENTIFIABLE, since $\text{dist}(3, 1) = 2 > 1$

Example

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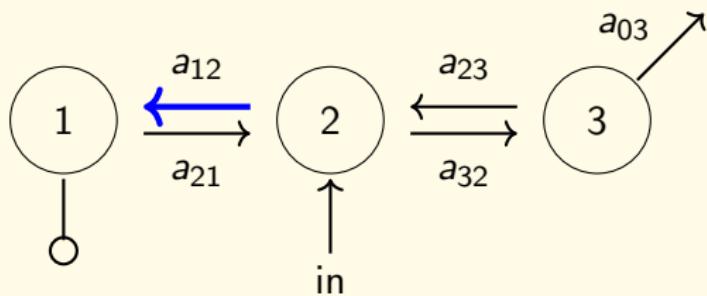


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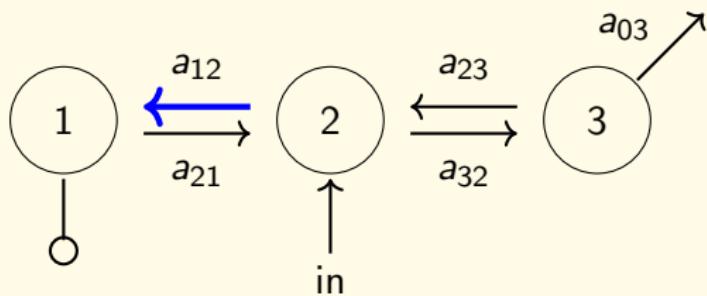
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A tree model $\mathcal{M} = (G, \{\text{in}\}, \{\text{out}\}, \text{Leak})$ is identifiable if and only if $\text{dist}(\text{in}, \text{out}) \leq 1$ and $|\text{Leak}| \leq 1$.

Example



IDENTIFIABLE, since $\text{dist}(2, 1) = 1 \leq 1$ and $|\text{Leak}| = 1 \leq 1$.

Conclusion

Theorem

For **ALL** linear compartmental models, we can generate defining input-output equations from the underlying graph.

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Theorem

For **tree models** with a single input and output, we completely classify *local* structural identifiability.

Remark

Biologists/modelers can use this information to design models which are structurally identifiable in the hope that they are practically identifiable.

Future Work

- generalize results on tree models to other linear compartmental models
- find more applications for new characterization of coefficients
 - consider *distinguishability*, i.e. the problem of determining whether two or more linear compartmental models fit a given set of measured data
 - look for patterns in the singular locus for *dividing edges*
 - consider identifiability versus observability relationship
- consider the problem of determining identifiability when multiple inputs/outputs are present

Acknowledgments and References

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Cashous Bortner, Elizabeth Gross, Nicolette Meshkat, Anne Shiu, and Seth Sullivant.

Identifiability of linear compartmental tree models and a general formula for the input-output equations.

Advances in Applied Mathematics, 146, May 2023.



Elizabeth Gross, Heather A. Harrington, Nicolette Meshkat, and Anne Shiu.

Linear compartmental models: input-output equations and operations that preserve identifiability.

SIAM J. Appl. Math., 79(4):1423–1447, 2019.

Thank you!!!

