

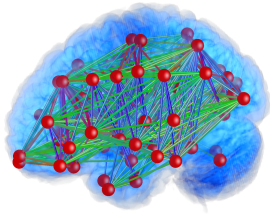
Local network identifiability: Path conditions

Antoine Legat and Julien M. Hendrickx

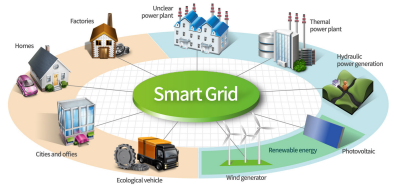
Benelux meeting 2023



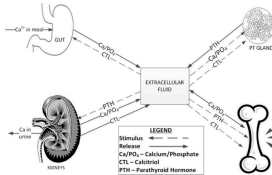
Neuroscience [Chiêm et al. 2021]



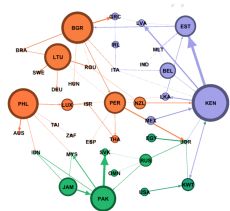
Power grids



Physiological models [Christie et al. 2014]



Stock market [Shahzad et al. 2018]



Excitations



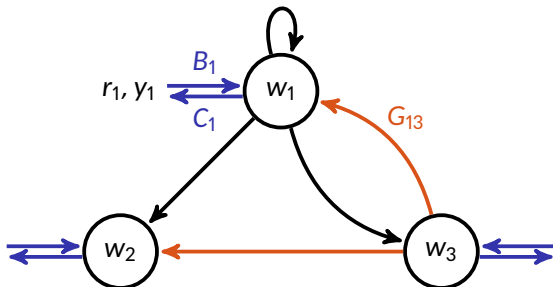
Measurements

Model

$$\text{state} \leftarrow w = G w + B r \rightarrow \text{excitation}$$

$$\text{measure} \leftarrow y = C w$$

From given exc/meas, can we recover the unknown transfer fcts?
i.e. From r at B and y at C , can we recover the unknown G_{ij} ?



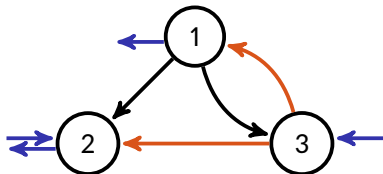
Assumptions: Network topology is known
Not all nodes are excited/measured
Global transfer matrix $\underbrace{C(I - G)^{-1}B}_{T(G)}$ is known

Example

state $\leftarrow w = G w + B r \rightarrow$ excitation

measure $\leftarrow y = C w$

From given exc/meas, can we recover the unknown transfer fcts?



Assumption: Global transfer matrix $\underbrace{C(I-G)^{-1}B}_{T(G)}$ is known

$$G = \begin{bmatrix} 0 & 0 & G_{13} \\ G_{21} & 0 & G_{23} \\ G_{31} & 0 & 0 \end{bmatrix} \quad C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

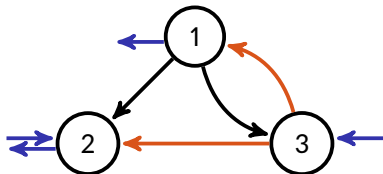
$$T \triangleq (I-G)^{-1} = \frac{1}{1-G_{13}G_{31}} \begin{bmatrix} 1 & 0 & G_{13} \\ G_{21} + G_{31}G_{23} & 1 & G_{23} + G_{13}G_{21} \\ G_{31} & 0 & 1 \end{bmatrix}$$

Example

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Assumption: Global transfer matrix $\underbrace{C(I-G)^{-1}B}_{T(G)}$ is known

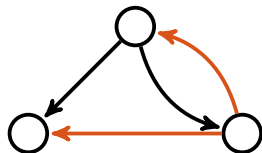
$$G = \begin{bmatrix} 0 & 0 & G_{13} \\ G_{21} & 0 & G_{23} \\ G_{31} & 0 & 0 \end{bmatrix} \quad C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$C T B = \frac{1}{1 - G_{13}G_{31}} \begin{bmatrix} 1 & 0 & G_{13} \\ G_{21} + G_{31}G_{23} & 1 & G_{23} + G_{13}G_{21} \\ G_{31} & 0 & 1 \end{bmatrix}$$

Outline

1. Context
2. Local identifiability
 - ▶ Algebraic condition
3. Decoupled identifiability
 - ▶ Path-based conditions
4. Future perspectives

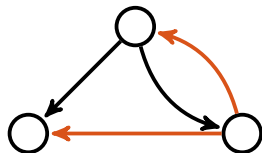
Identifiability



Network *identifiable* if the unknown transfer fcts can be recovered.

Does $\underbrace{C T B}_{\text{known}} = C (I - G)^{-1} B$ admit a unique solution G ?

Identifiability is generic



Network *identifiable* if the unknown transfer fcts can be recovered.

Does $\underbrace{C T B}_{\text{known}} = C (I - G)^{-1} B$ admit a unique solution G ?

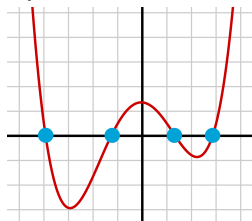
Identifiability is a property of the **graph topology**.

It does almost **not** depend on parameters G_{ij} .

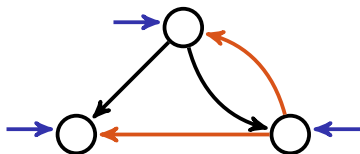
Identifiability is a **generic** property. It holds:

- ▶ either for almost all parameters G_{ij} ;
- ▶ or for no parameters G_{ij} .

Example: the nonzeroness of a polynomial is **generic**



State of the Art



Network *identifiable* if the unknown transfer fcts can be recovered.

Does $\underbrace{C T B}_{\text{known}} = C (I - G)^{-1} B$ admit a unique solution G ?

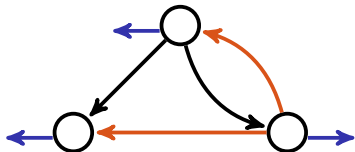
All nodes excited: Necessary and sufficient path-based condition

i.e. $B = I$

[Hendrickx, Gevers, Bazanella 2017]

Algo allocating measurements in the graph

[Cheng, Shi, Van den Hof 2019]



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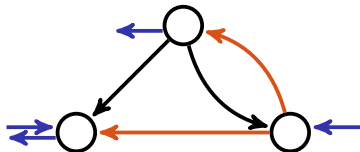
[Hendrickx, Gevers, Bazanella 2017]

Also allocating measurements in the graph

[Cheng, Shi, Van den Hof 2019]

All nodes measured: Dual results

i.e. $C = I$



Network *identifiable* if the unknown transfer fcts can be recovered.

Does $\underbrace{C T B}_{\text{known}} = C (I - G)^{-1} B$ admit a unique solution G ?

All nodes excited: Necessary and sufficient path-based condition
i.e. $B = I$ [Hendrickx, Gevers, Bazanella 2017]

Algo allocating measurements in the graph

[Cheng, Shi, Van den Hof 2019]

All nodes measured: Dual results
i.e. $C = I$

General case: Need to linearize

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From the definition of global identifiability...

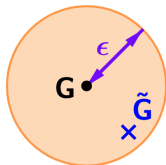
Definition: Network globally identifiable at G if for all \tilde{G} :

$$C T(\tilde{G}) B = C T(G) B \Rightarrow \tilde{G} = G$$

... we introduce *local* identifiability

Definition: Network *locally* identifiable at G if for all \tilde{G} on an ϵ -ball:

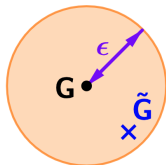
$$C T(\tilde{G}) B = C T(G) B \Rightarrow \tilde{G} = G$$



Local identifiability is necessary

Definition: Network *locally* identifiable at G if for all \tilde{G} on an ϵ -ball:

$$C T(\tilde{G}) B = C T(G) B \Rightarrow \tilde{G} = G$$

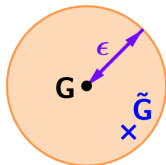


- ▶ Necessary for global identifiability
- ▶ No counter-example to sufficiency known

Algebraic condition

Definition: Network *locally* identifiable at G if for all \tilde{G} on an ϵ -ball:

$$CT(\tilde{G})B = CT(G)B \Rightarrow \tilde{G} = G$$

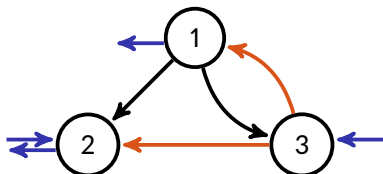


- ▶ Necessary for global identifiability
- ▶ No counter-example to sufficiency known

Theorem 1 (CDC 2020)

G generically locally identif $\Leftrightarrow \boxed{CT\Delta TB = 0 \Rightarrow \Delta = 0} \quad \forall \Delta$
almost everywhere

Example



$$G = \begin{bmatrix} 0 & 0 & G_{13} \\ G_{21} & 0 & G_{23} \\ G_{31} & 0 & 0 \end{bmatrix}$$

$$C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$T \triangleq (I - G)^{-1} = \frac{1}{\underbrace{1 - G_{13}G_{31}}_{\triangleq D}} \begin{bmatrix} 1 & 0 & G_{13} \\ G_{21} + G_{31}G_{23} & 1 & G_{23} + G_{13}G_{21} \\ G_{31} & 0 & 1 \end{bmatrix}$$

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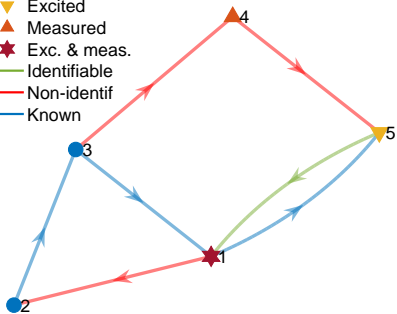
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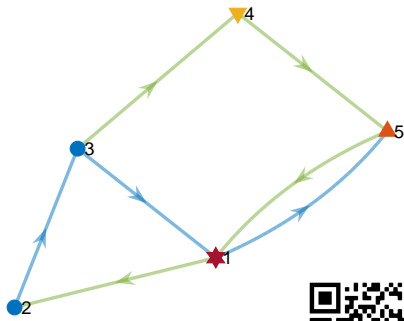
almost everywhere

→ Probability-1 algorithm: randomized, proba 0 of inaccuracy

- ▼ Excited
- ▲ Measured
- ★ Exc. & meas.
- Identifiable
- Non-identif
- Known



(a)



(b)



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What we have so far

Global identifiability

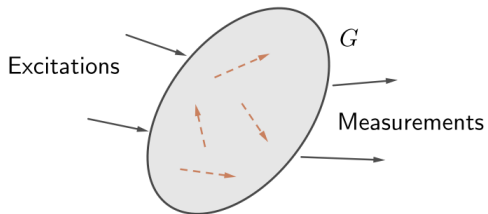
$$C\tilde{T}B = CTB \Rightarrow \tilde{G} = G$$



Local identifiability

$$CT\Delta TB = 0 \Rightarrow \Delta = 0$$

→ *Graph interpretation?*



Interpretation on a larger network

Global identifiability

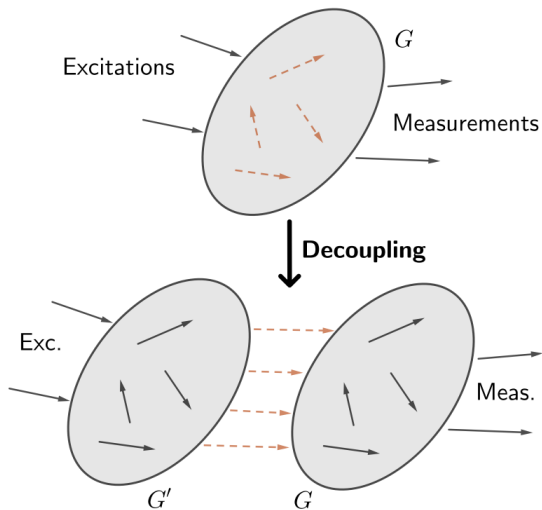
$$C\tilde{T}B = CTB \Rightarrow \tilde{G} = G$$



Local identifiability

$$CT\Delta TB = 0 \Rightarrow \Delta = 0$$

Left graph = right graph?



Decoupled identifiability

Global identifiability

$$C\tilde{T}B = CTB \Rightarrow \tilde{G} = G$$



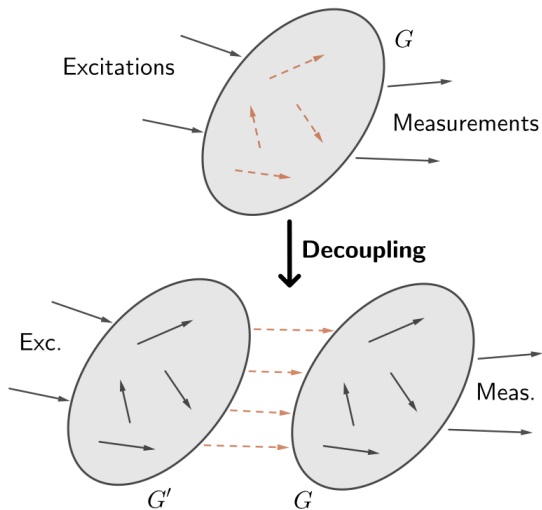
Local identifiability

$$CT\Delta TB = 0 \Rightarrow \Delta = 0$$



Decoupled identifiability

$$CT\Delta T'B = 0 \Rightarrow \Delta = 0$$



Necessary... and sufficient?

Global identifiability

$$C\tilde{T}B = CTB \Rightarrow \tilde{G} = G$$

↓ ↑?

Local identifiability

$$CT\Delta TB = 0 \Rightarrow \Delta = 0$$

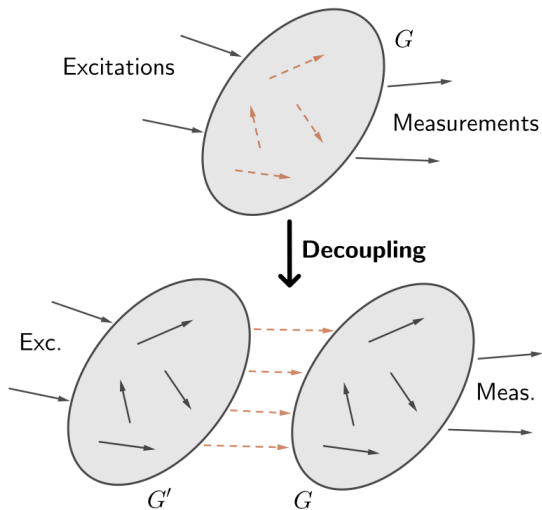
↓ ↑?

Decoupled identifiability

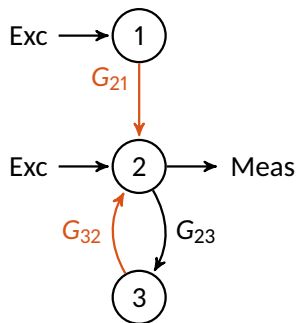
$$CT\Delta T'B = 0 \Rightarrow \Delta = 0$$

No counter-example known

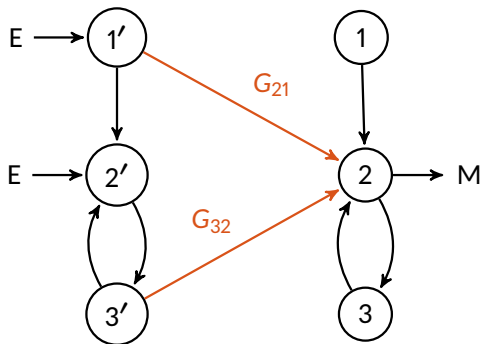
github.com/alegat/identifiable



Basic example



(a)



(b)

(b) is the decoupled network of (a)

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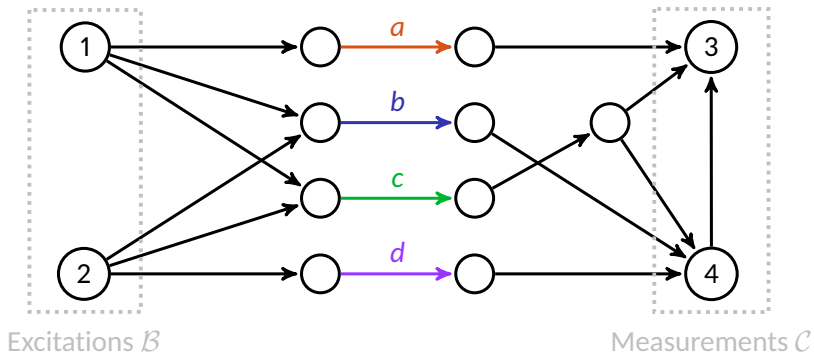
Theorem 1 - CDC 2020

G generically locally identif \Leftrightarrow

$$\boxed{CT\Delta TB = 0 \Rightarrow \Delta = 0} \quad \forall \Delta$$

almost everywhere

Example:

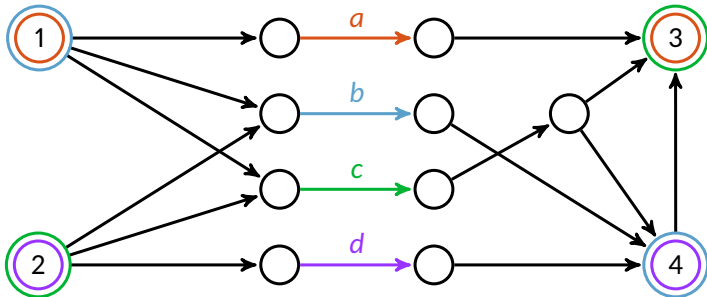


Theorem 2 (weak condition) – CDC 2021

Decoupled identif $\Rightarrow \exists$ a connected bijective assignment
 \exists *only one* connected bijective assignment \Rightarrow Decoupled identif

Example of assignment:

$a \rightarrow (1, 4)$, $b \rightarrow (1, 3)$, $c \rightarrow (2, 3)$, $d \rightarrow (2, 4)$



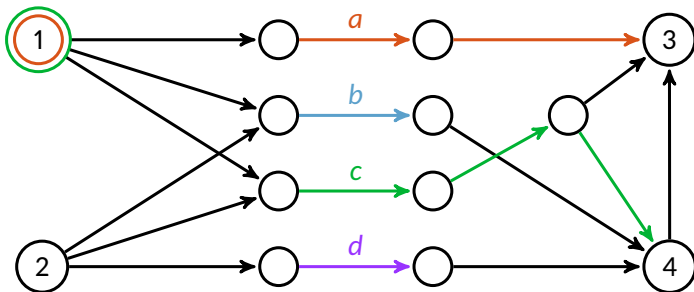
Theorem 3 (strong condition) - CDC 2021

Decoupled identif $\Rightarrow \exists$ a connected assignment s.t:

- ▶ for each excitation i , there are 2 vertex-disjoint paths between the edges assigned to i and the measurements
- ▶ dual condition for the measurements

If \exists only one such \Rightarrow Decoupled identif

Example: Excitation 1: **OK**



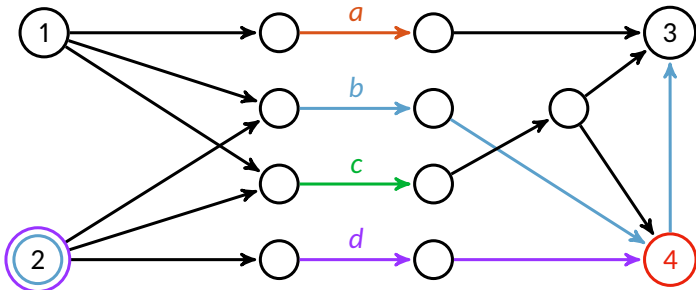
Theorem 3 (strong condition)

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- ▶ for each excitation i , there are 2 vertex-disjoint paths between the edges assigned to i and the measurements
- ▶ dual condition for the measurements

If \exists only one such \Rightarrow Decoupled identif

Example: Excitation 2 : KO



Theorem 2 (weak condition)

Decoupled identif $\Rightarrow \exists$ a connected *bijjective* assignation
 \exists *only one* connected *bijjective* assignation \Rightarrow Decoupled identif

Theorem 3 (strong condition)

Decoupled identif $\Rightarrow \exists$ a connected *vertex-disjoint* assignation
 \exists *only one* connected *v-disjoint* assignation \Rightarrow Decoupled identif

→ There *could be a stronger condition* combining Theorems 2 & 3, extending previous results under full excitation/measurement

[Hendrickx, Gevers, Bazanella 2017]

Take-home message

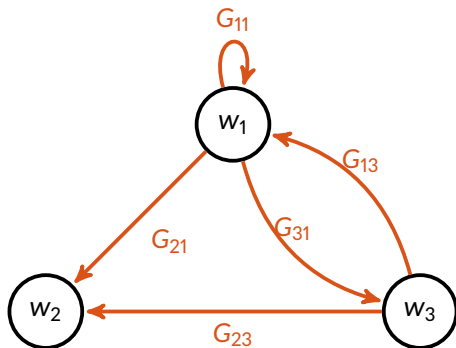
- ▶ Introduced *decoupled* identifiability
 - ▶ Necessary for local and global identifiability
 - ▶ New: larger graph which decouples excitations and measurements
- ▶ Derived *path-based necessary conditions* which also apply to local and global identifiability
- ▶ Whether the sufficient conditions extend as well remains an open question
- ▶ There *could be a stronger version* of our conditions, extending previous results under full excitation/measurement
- ▶ **Further work:** when not all edges are identifiable, obtain a path-based condition for the recovery of some edges

Back-up slides

Model

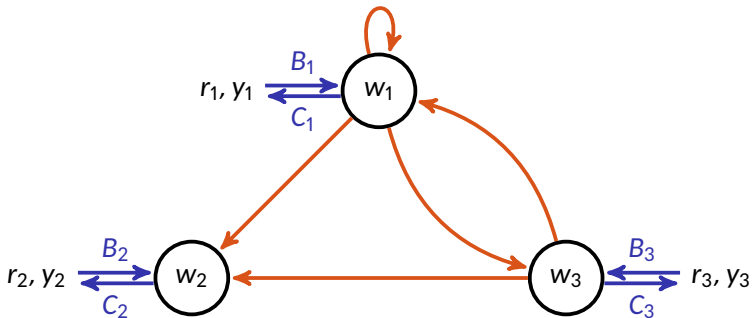
$$\text{state} \leftarrow w_i(t) = \sum G_{ij}(q) w_j(t)$$

q is the shift operator, i.e. $q^{-1}w(t) = w(t-1)$



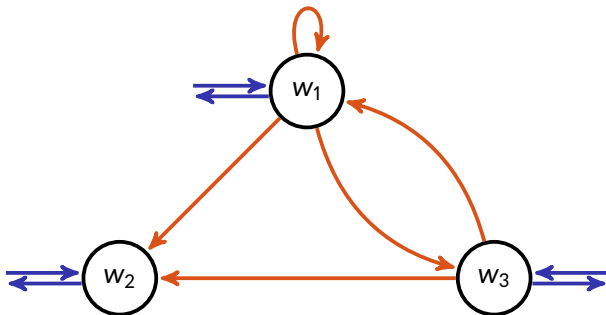
state $\leftarrow w_i(t) = \sum G_{ij}(q) w_j(t) + B_i r_i(t) \rightarrow$ excitation

measure $\leftarrow y_i(t) = C_i w_i(t) \quad B_i, C_i \in \{0, 1\}$



$$\text{state} \leftarrow w = G w + B r \rightarrow \text{excitation}$$
$$\text{measure} \leftarrow y = C w$$

Which nodes to excite/measure to recover the transfer functions?
i.e. how to choose B, C to accurately recover G ?

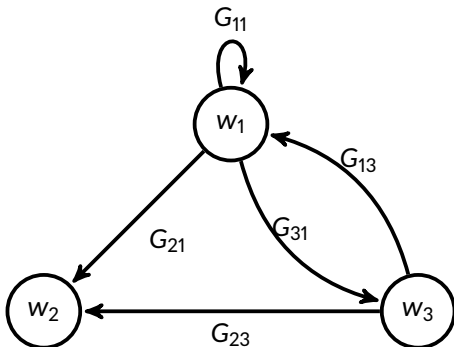


Transfer functions that can be recovered are *identifiable*

Network topology is defined by the nonzero entries of G , and is assumed to be known (often the case).

$$G = \begin{bmatrix} G_{11} & 0 & G_{13} \\ G_{21} & 0 & G_{23} \\ G_{31} & 0 & 0 \end{bmatrix}$$

Theorem: *Identifiability* is a generic property of *network topology*: it only depends* on the structure of G , but not on its parameters G_{ij} .



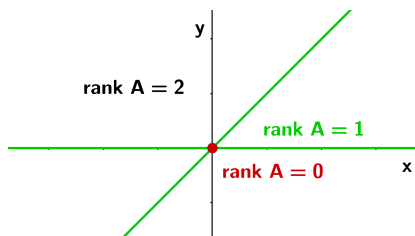
Genericity

- ▶ A generic property holds everywhere except possibly on a *lower-dimensional set*.
- ▶ A lower-dimensional set has *Lebesgue-measure zero*
→ 0-probability of falling in this set when sampling randomly

Example: The matrix

$$A = \begin{bmatrix} x & 0 \\ 0 & x - y \end{bmatrix}$$

has generic rank 2. Its rank drops on $\{x = 0\} \cup \{x = y\}$.



Identifiability is generic – example

Global input-output transfer function:

$$CTB \triangleq C(I - G)^{-1}B = \begin{pmatrix} G_{42}G_{21} + G_{43}G_{31} & G_{42} & G_{43} & 1 & 0 \\ G_{52}G_{21} + G_{53}G_{31} & G_{52} & G_{53} & 0 & 1 \end{pmatrix}$$
$$\Rightarrow G_{42}, G_{43}, G_{52}, G_{53} \text{ identif, and } \begin{pmatrix} G_{42} & G_{43} \\ G_{52} & G_{53} \end{pmatrix} \begin{pmatrix} G_{21} \\ G_{31} \end{pmatrix} = \begin{pmatrix} T_{41} \\ T_{51} \end{pmatrix}$$

$\Rightarrow G_{21}, G_{31}$ identifiable except when $G_{42}G_{53} + G_{43}G_{52} = 0$.

