

Local dynamics identification via a graph-theoretical approach

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1. Introduction
 - 1.1 Motivations
 - 1.2 Model
2. Previous results
3. General case
4. Algorithm
5. Conclusion

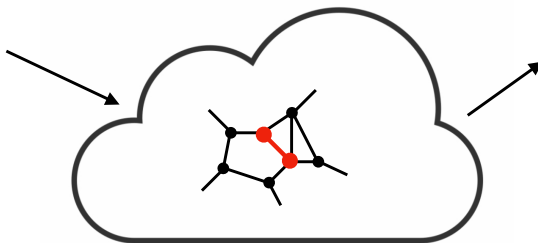
Long-term application



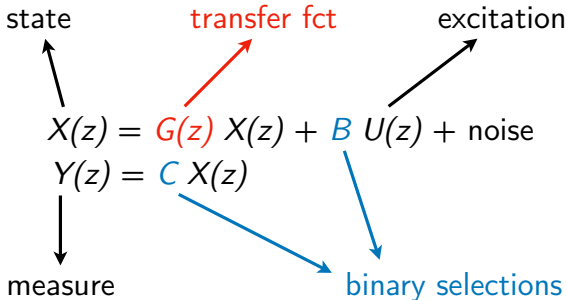
Case study (my Master's thesis)



Excitation

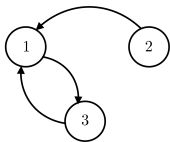


Measurement



Which nodes to excite/measure to recover the **transfer functions**?

Theorem: Identifiability is a *generic* property of network topology.

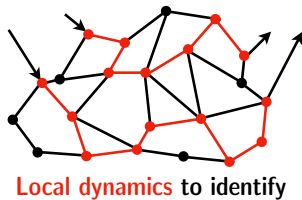
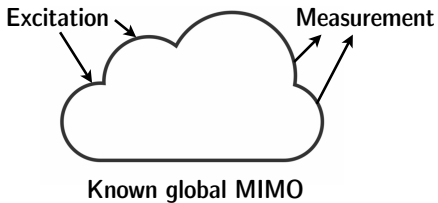


$$G = \begin{bmatrix} 0 & G_{12} & G_{13} \\ 0 & 0 & 0 \\ G_{31} & 0 & 0 \end{bmatrix}$$

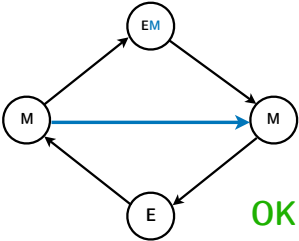
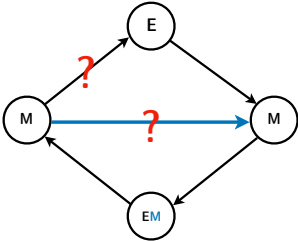
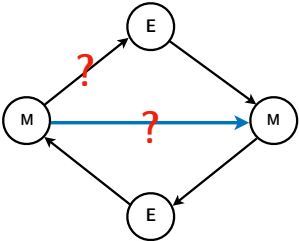
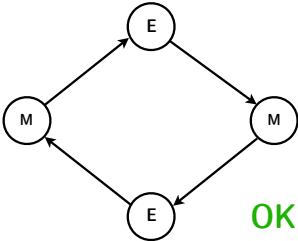
Assumption: Topology is known (often the case).

Approach

- **Global** graph theoretical approach
- Local dynamics from global input-output



Motivating example

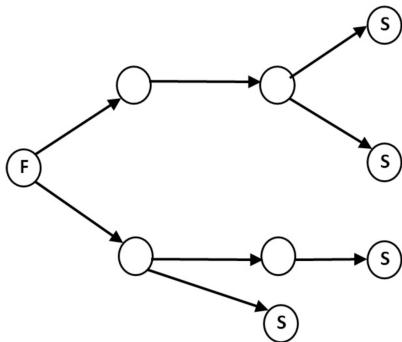


1. Introduction
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 - 2.1 Trees
 - 2.2 Full excitation/measurement
 - 2.3 Pseudotrees
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Trees

Theorem (Bazanella et al. 2019)

A tree is generically identifiable IFF all fountains are excited, all sinks are measured and all nodes are either excited or measured.

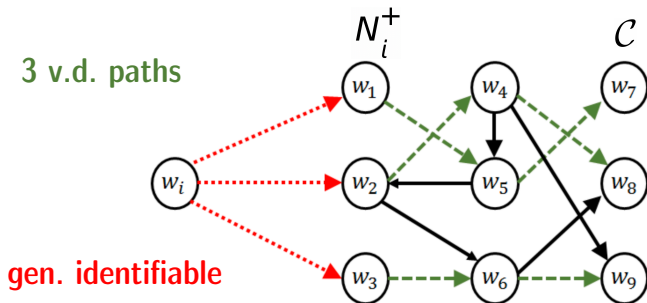


Definition: Fountains: no incoming edge – Sinks: no outgoing edge

Full excitation/measurement

Theorem (Full excitation – Hendrickx et al. 2018)

All transfer functions leaving node i are generically identifiable IFF there are $|N_i^+|$ vertex disjoint paths from N_i^+ to \mathcal{C} .



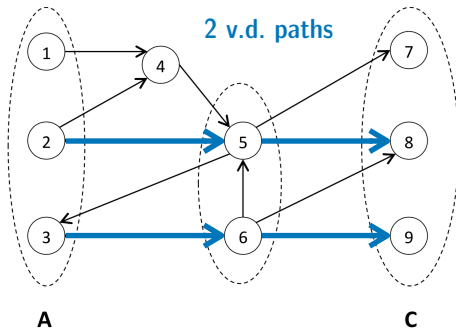
Dual (full measurement): All TFs arriving at node j are generically identifiable IFF there are $|N_j^-|$ vertex disjoint paths from N_j^- to \mathcal{B} .

Vertex disjoint paths

Theorem (Hendrickx et al. 2018)

- $b_{\mathcal{A} \rightarrow \mathcal{C}}$: max number of vertex disjoint paths
- $T_{\mathcal{C}, \mathcal{A}}(G)$: restriction of $T(G) \triangleq (I - G)^{-1}$ to nodes \mathcal{C} and \mathcal{A}

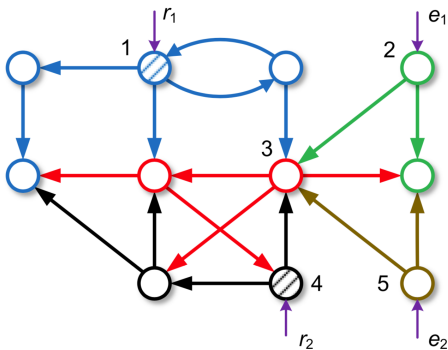
$$b_{\mathcal{A} \rightarrow \mathcal{C}} = \text{generic rank}(T_{\mathcal{C}, \mathcal{A}}(G))$$



Pseudotrees

Theorem (Full measurement - Cheng, Shi, Van den Hof 2019)

The network is generically identifiable IFF there exists a disjoint pseudotree covering with excited roots (+ dual for full excitation).



Analogy: Vertex disjoint paths - Pseudotree covering

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 - 3.1 Local generic identifiability
 - 3.2 Whole network
 - 3.3 Specific edge
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Local generic identifiability

Definition

The transfer function $G_{ji}(z)$ is **locally** generically identifiable from \mathcal{B} & \mathcal{C} if, for any parametrization $G(P, z)$, there exists $\tilde{G}(z)$ consistent with the graph, $\epsilon > 0$ such that $\|\tilde{G}(z) - G(P, z)\| < \epsilon$, and there holds

$$C(I - G(P, z))^{-1}B = C(I - \tilde{G}(z))^{-1}B \Rightarrow G_{ji}(P, z) = \tilde{G}_{ji}(z) \quad (1)$$

\forall parameters P except possibly those lying on a zero measure set.

One can linearize (1) to obtain

$$CT\Delta TB = 0 \Rightarrow \Delta_{ji} = 0,$$

where $T \triangleq (I - G(z))^{-1}$ and Δ is consistent with the graph.

Whole network

For the whole network, it becomes

$$CT\Delta TB = 0 \Rightarrow \Delta_{j\neq i} = 0,$$

which can be reformulated as a linear system

$$K\delta = 0 \Rightarrow \delta = 0,$$

where K is the restriction of $B^T T^T \otimes CT$ to the columns corresponding to actual edges, $\delta \in \mathbb{R}^{|E|}$, and $|E|$ denotes the number of edges.

Theorem

G is locally generically identifiable IFF generic rank $K = |E|$.

Specific edge

Otherwise, $\ker K \neq \emptyset \longrightarrow$ inspect $\ker K$ to find problematic edges
The linearization

$$CT\Delta TB = 0 \Rightarrow \Delta_{ji} = 0$$

can be reformulated as

$$K\delta = 0 \Rightarrow \delta_e = 0,$$

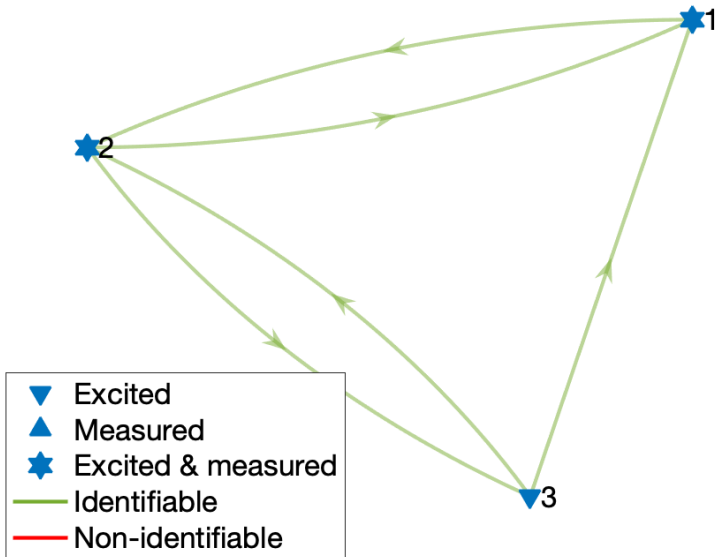
where e corresponds to edge (ji) .

Theorem

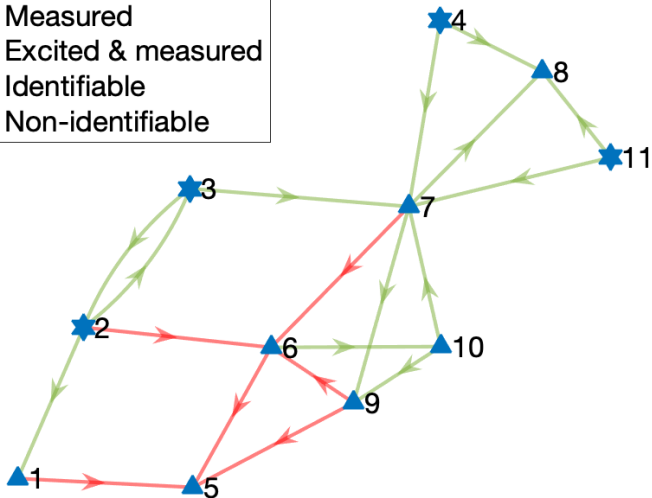
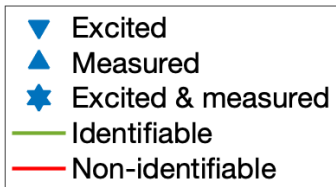
The transfer function e is locally generically identifiable IFF for each $\delta \in \ker K$, there holds $\delta_e = 0$.

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Symbolic run



Stochastic run (probability 1)



Conclusion

- Introduced local generic identifiability
 - By a linearization, obtained necessary and sufficient conditions for recovering (i) the whole network (ii) a specific edge
 - Algorithm: symbolic and exact, or stochastic and efficient
 - Future work
 - Characterize our results in graph-theoretical terms
 - Extend our results to nonlocal generic identifiability
- Hendrickx, J.M., Gevers, M., and Bazanella, A.S. (2018). Identifiability of dynamical networks with partial node measurements. *IEEE Transactions on Automatic Control*.
- Bazanella, A. S., Gevers, M., and Hendrickx, J. (2019). Network Identification with Partial Excitation and Measurement. In *IEEE Conference on Decision and Control. Proceedings*.
- Cheng, X., Shi, S., and Van den Hof, P. M. (2019). Allocation of excitation signals for generic identifiability of dynamic networks. In *IEEE Conference on Decision and Control. Proceedings*.