

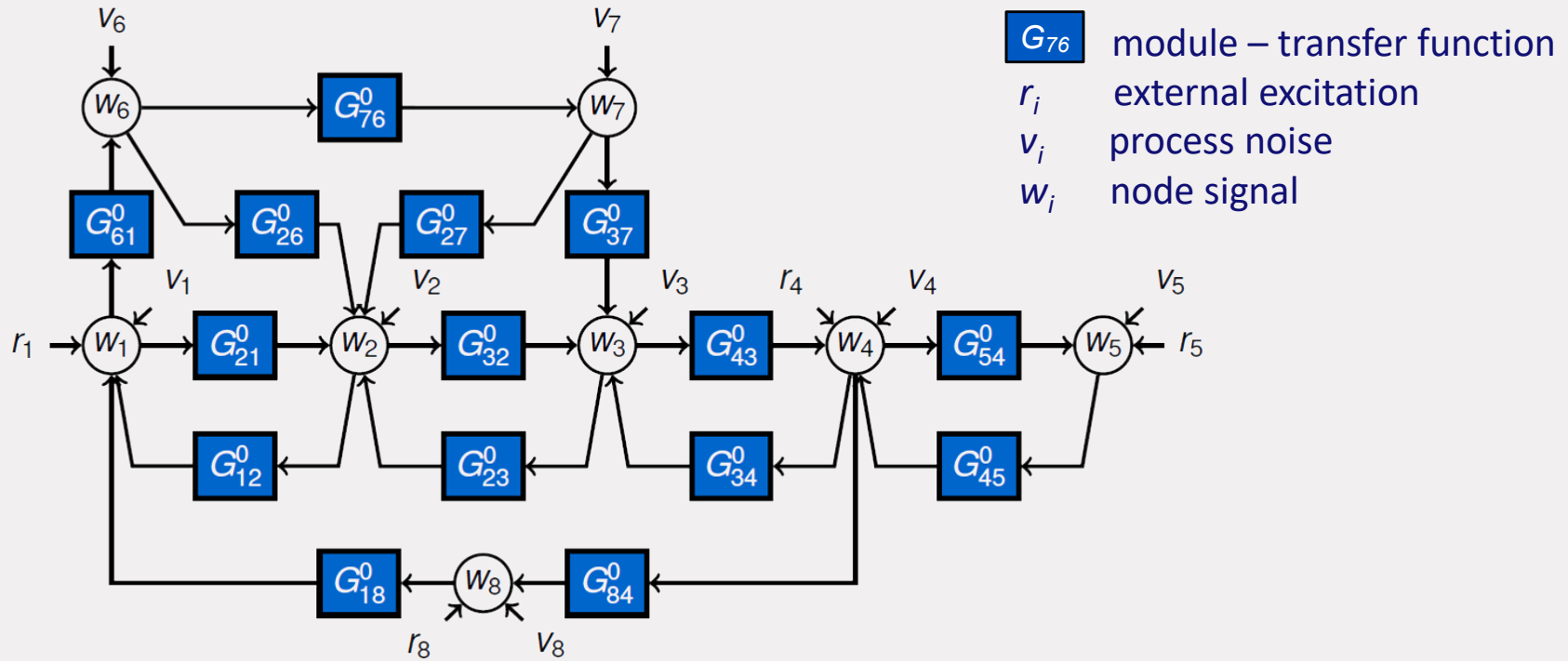
Excitation allocation for generic identifiability of linear dynamic networks with fixed modules

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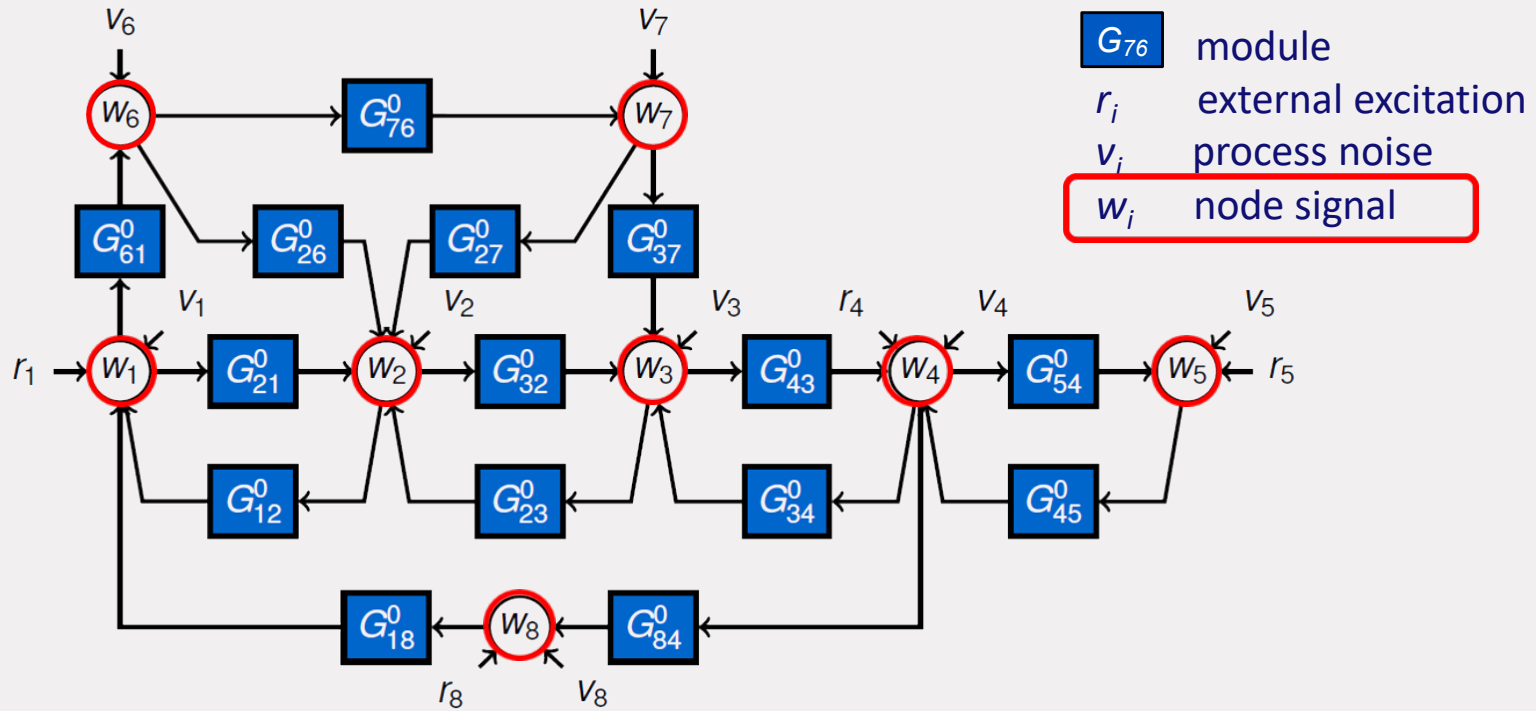
CDC 2022, 7 December 2022, Cancun, Mexico

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Dynamic network



Dynamic network setup



Dynamic network setup

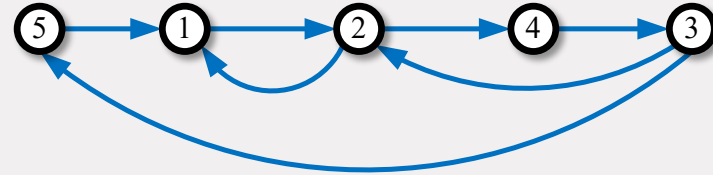
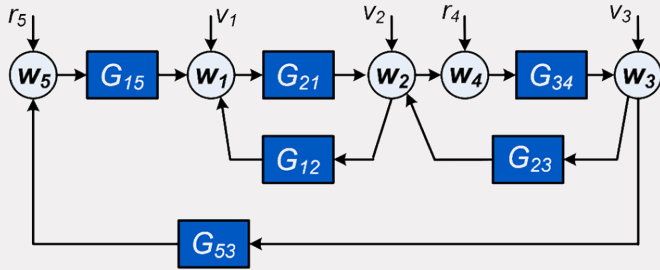
Collecting all equations:

$$\begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_L \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & G_{12}^0 & \cdots & G_{1L}^0 \\ G_{21}^0 & 0 & \cdots & G_{2L}^0 \\ \vdots & \cdots & \cdots & \vdots \\ G_{L1}^0 & G_{L2}^0 & \cdots & 0 \end{bmatrix}}_{\text{Network matrix } G^0(q)} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_L \end{bmatrix} + R^0 \begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_K \end{bmatrix} + H^0 \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_p \end{bmatrix}$$

$$w(t) = G^0(q)w(t) + R^0(q)r(t) + v(t); \quad v(t) = H^0(q)e(t); \quad \text{cov}(e) = \Lambda$$

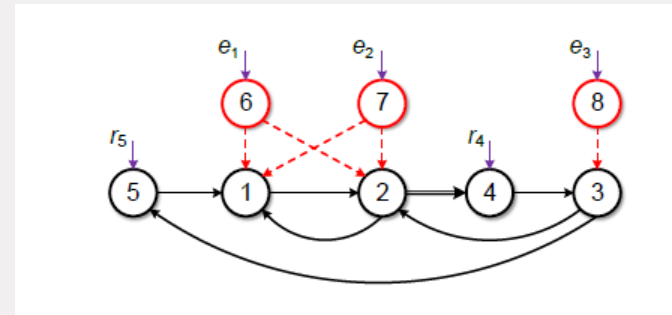
- Typically R^0 is just a (static) selection matrix, indicating which nodes have an excitation signal.
- The topology of the network is encoded in the structure (non-zero entries) of G^0 .
- r and e are called **external signals**.

Dynamic network setup – directed graph



Nodes are vertices; modules/links are edges

Extended graph:
including the external signals
and disturbance correlations



Network identifiability

The identifiability problem:

The network model:

$$w(t) = G(q)w(t) + R(q)r(t) + \underbrace{H(q)e(t)}_{v(t)}$$

can be transformed with any rational $P(q)$:

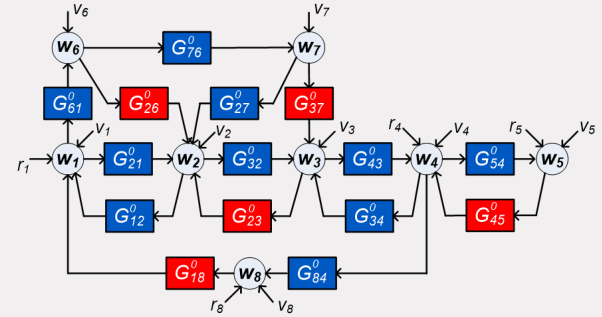
$$P(q)w(t) = P(q)\{G(q)w(t) + R(q)r(t) + H(q)e(t)\}$$

$$w(t) = (I - P(q))w(t) + P(q)\{G(q)w(t) + R(q)r(t) + H(q)e(t)\}$$

to an equivalent model:

$$w(t) = \tilde{G}(q)w(t) + \tilde{R}(q)r(t) + \tilde{H}(q)e(t)$$

➔ **Nonuniqueness**, unless there are structural constraints on G, R, H .



[1] Weerts, Linder et al., *Automatica*, 2019.

[2] Bottegal et al., *SYSID* 2018

Network identifiability

Consider a **network model set**:

$$\mathcal{M} = \{(G(\theta), R(\theta), H(\theta))\}_{\theta \in \Theta}$$

representing structural constraints on the considered models:

- modules that are fixed and/or zero (topology)
- locations of excitation signals
- disturbance correlation

Generic identifiability of \mathcal{M} :

- There do not exist distinct equivalent models (generating the same data)
- for almost all models in the set.

Full measurement set-up: all w and r signals are assumed to be measured.

[1] Weerts et al., Automatica, March 2018;

[2] Bazanella, CDC2017; Hendrickx et al., IEEE-TAC, 2019.

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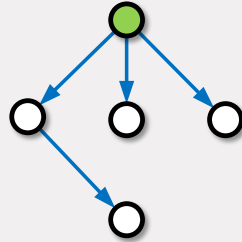
- Dynamic networks and identifiability
- **Current solution with pseudotree covering**
- Limitation of the current solution
- Extension with SIMUGs to reduce conservatism

Synthesis solution for network identifiability

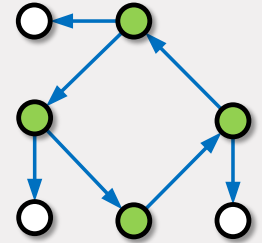
Allocating external signals for **generic identifiability**:

1. Define **Pseudo-tree**: a directed network where each vertex has indegree ≤ 1

Tree with root in green



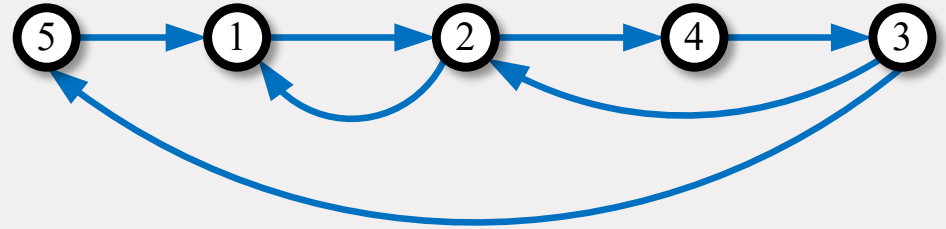
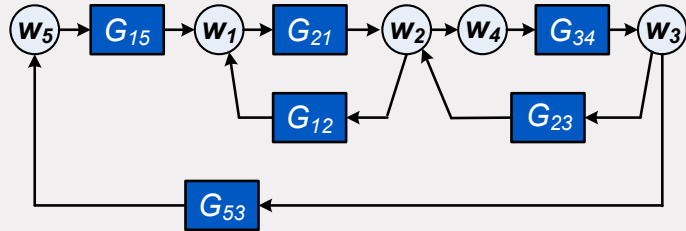
Cycle with outgoing trees;
Any node in cycle is root



2. Cover the network (model set) graph with **edge-disjoint** pseudo-trees, where all out-neighbors of a vertex are in the same pseudo-tree
3. Assign an independent external signal (r or e) at a root of each pseudo-tree.

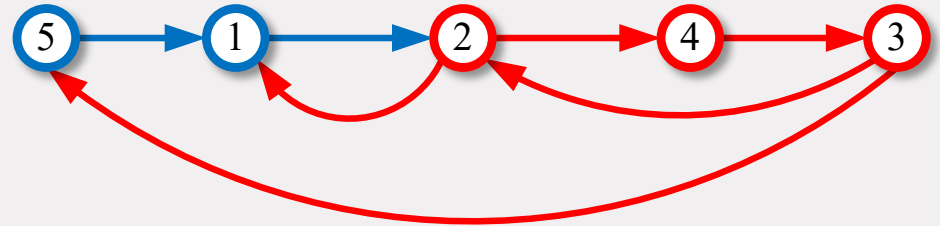
This guarantees **generic identifiability** of the model set.

Where to allocate external excitations for network identifiability?

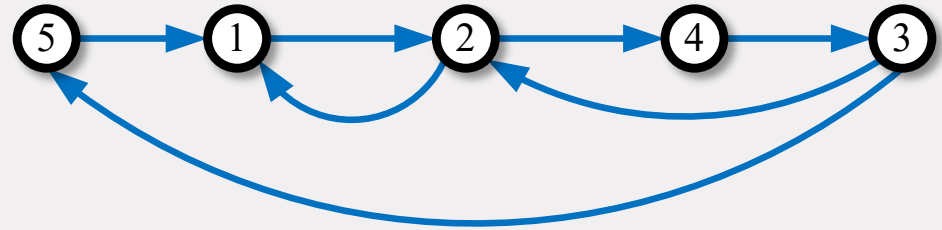
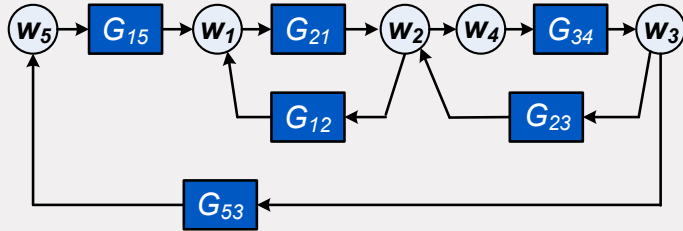


All indicated modules are parametrized

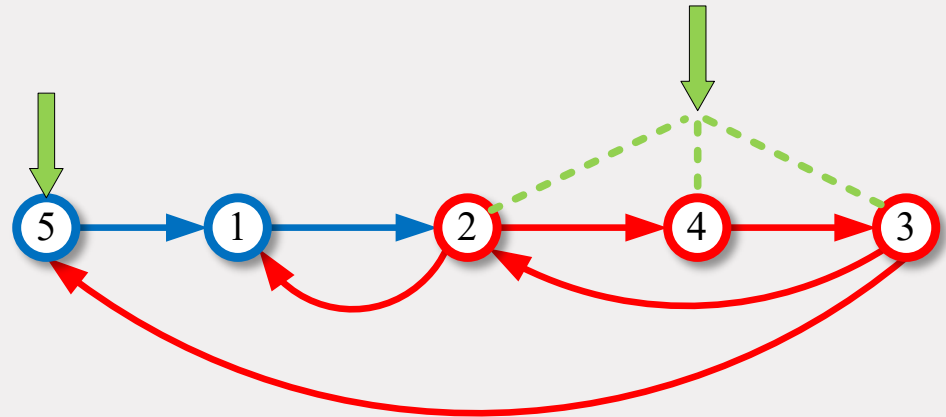
Two disjoint pseudo-trees



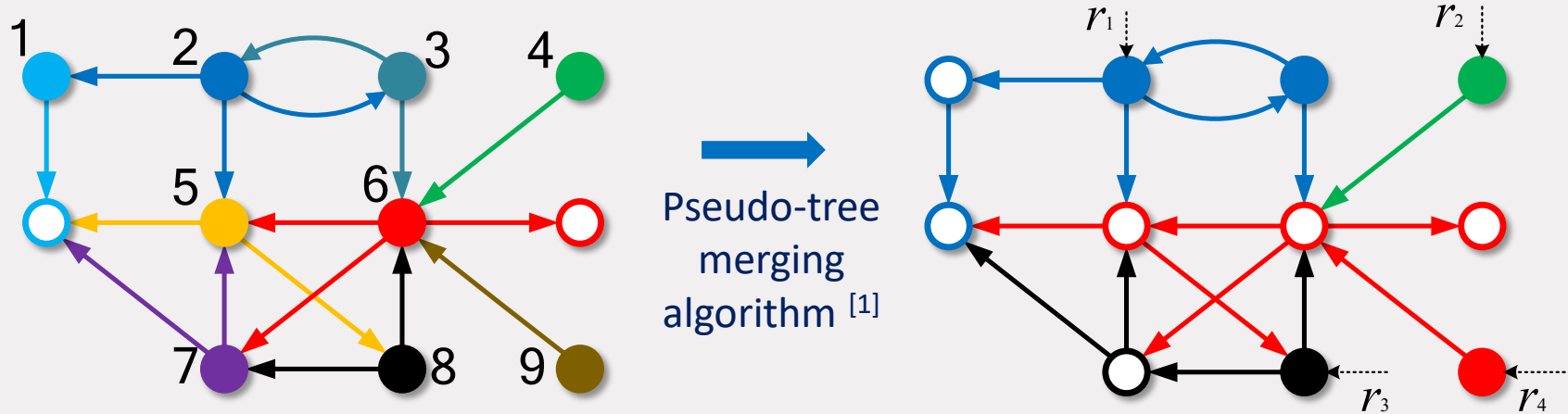
Where to allocate external excitations for network identifiability?



Two independent excitations
guarantee
generic network identifiability

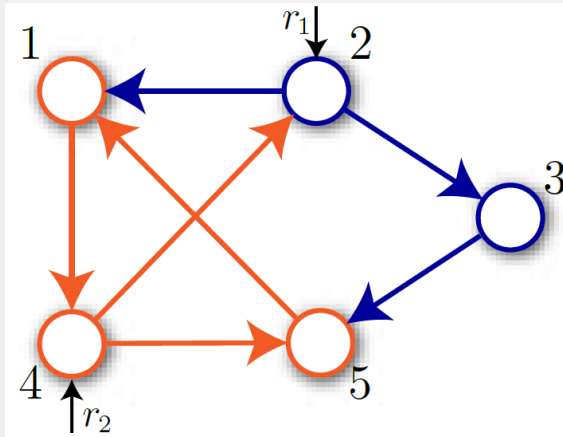


Finding the pseudo-trees: merging algorithm

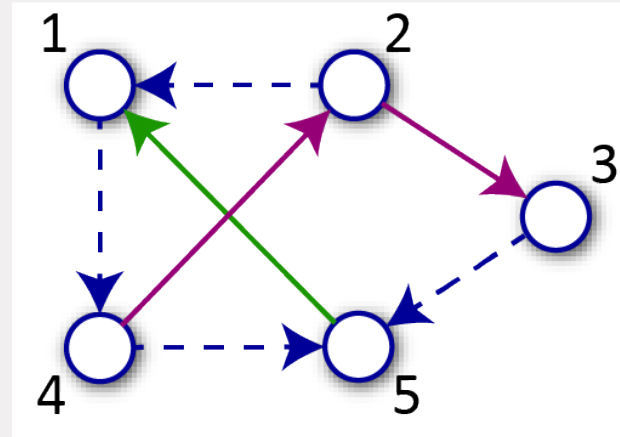


- Merging is implemented through an algebraic procedure
- Known (fixed) edges do not need to be covered

Conservatism of current solution

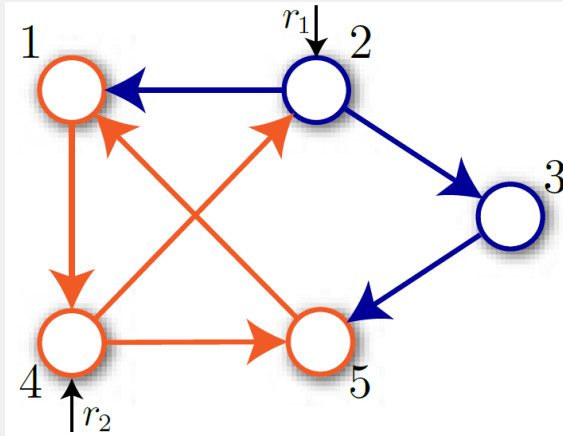


2 pseudo-trees for all edges
parametrized

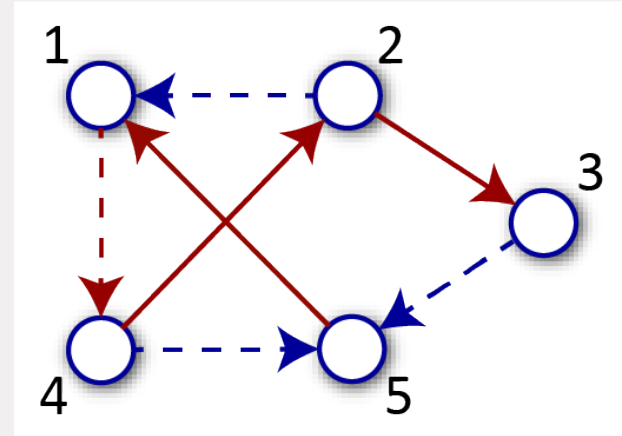


Dashed edges are fixed.
Parametrized edges covered
by 2 pseudo-trees (4-2-3, 5-1)

Conservatism of current solution



2 pseudo-trees for all edges
parametrized



Dashed edges are fixed.
Parametrized edges covered
by 2 pseudo-trees (4-2-3, 5-1)

But: all parametrized edges
can be covered by 1 pseudo-
tree: (5-1-4-2-3)

New approach: explicit incorporation of fixed edges

Define **multi-rooted graph**: directed graph for which there is a nonempty set of roots from all of which there exist paths to every vertex in the graph.

Define **single source identifiable multi-rooted graph (SIMUG)**: multi-rooted graph where each vertex has an indegree of *parametrized edges* ≤ 1 .

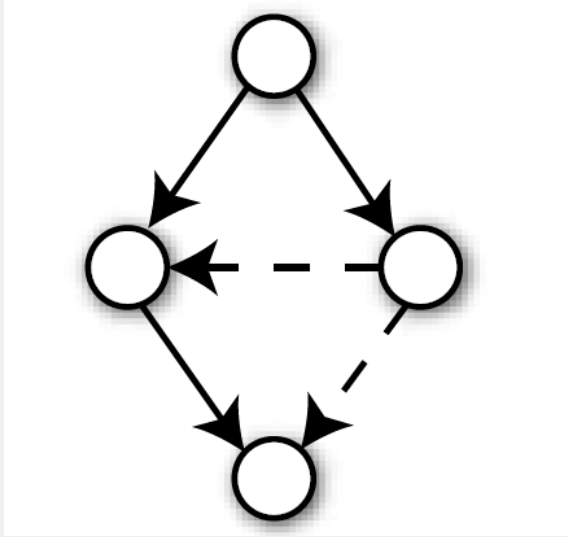
Define **edge-disjoint SIMUGs**: no common edges and for each vertex all outgoing edges are in the same SIMUG.

Result:

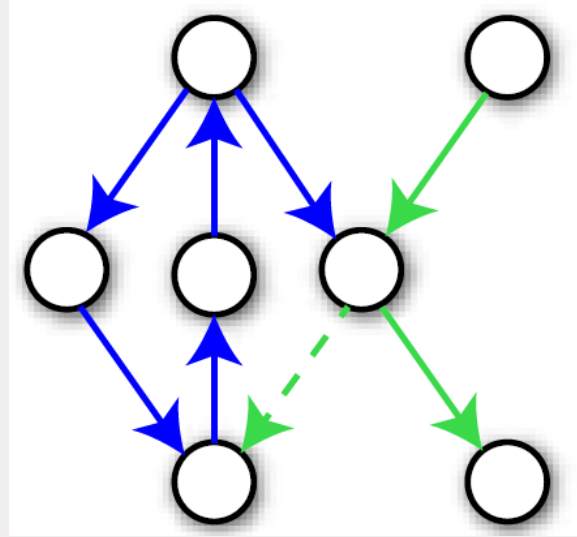
Generic network identifiability holds if:

- All *parametrized edges* in the network graph are covered by edge-disjoint SIMUGs, and
- An external signal (r or e) is applied to one vertex in the root set of every SIMUG.

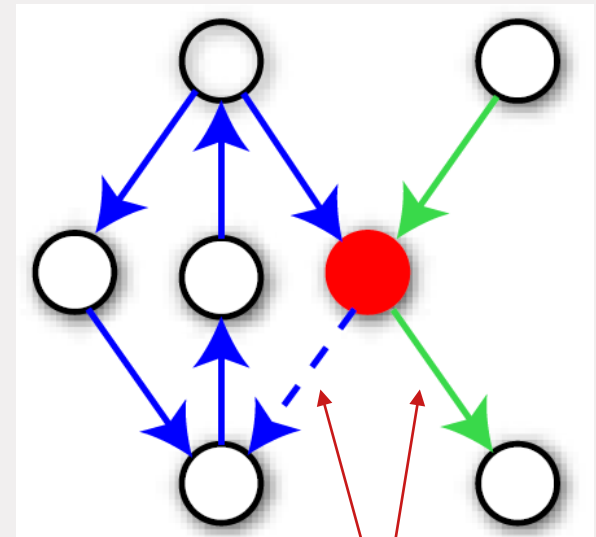
Examples of SIMUGs



One SIMUG



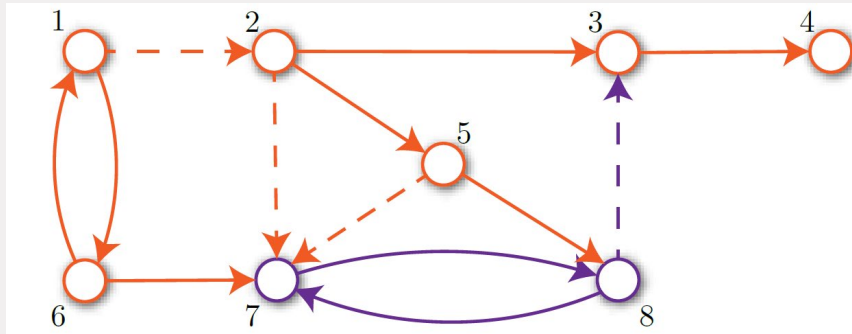
Two edge-disjoint SIMUGs



Two SIMUGs,
but not edge-disjoint

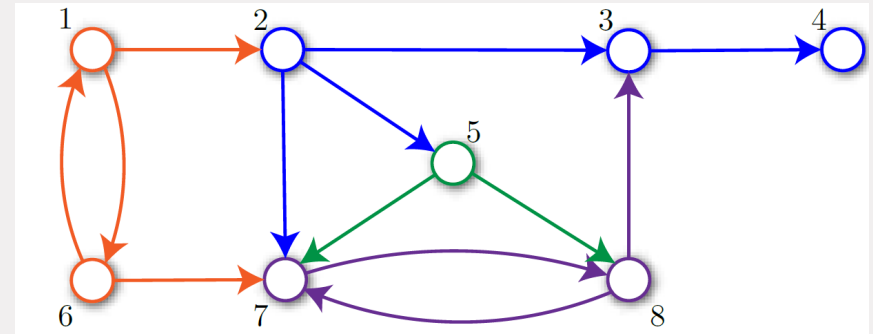
8-node example with fixed links:

Dashed links are fixed (known modules)



Network graph covered by 2 edge-disjoint SIMUGs

Two external signals required for identifiability: (1 or 6) and (7 or 8)

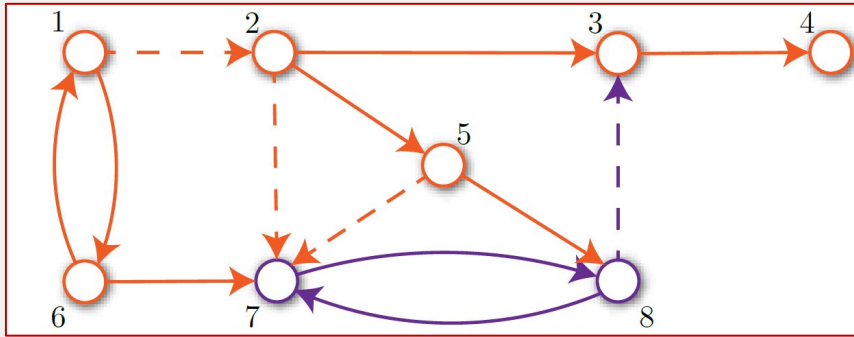


When discarding the fixed-property of links:
Network graph covered by 4 edge-disjoint pseudo-trees

Four external signals required for identifiability: (1 or 6), (7 or 8), 2 and 5

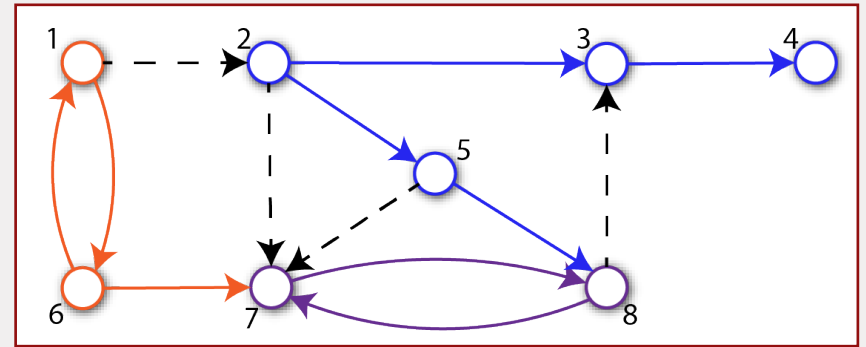
8-node example with fixed links:

Dashed links are fixed (known modules)



Network graph covered by 2 edge-disjoint SIMUGs

Two external signals required for identifiability: (1 or 6) and (7 or 8)



When applying pseudo-tree covering of the parametrized edges:

graph covered by 3 pseudo-trees

Three external signals required for identifiability: (1 or 6), (7 or 8), and 2.

Merging algorithm

Consider the disjoint SIMUG covering: $\{\mathcal{T}_1, \dots, \mathcal{T}_n\}$

Construct the mergability matrix \mathfrak{M} with

$$\mathfrak{M}_{ij} = \begin{cases} 1 & \text{if } \mathcal{T}_i \text{ is mergeable to } \mathcal{T}_j \\ \emptyset & \text{if there are no vertices in } \mathcal{T}_i \cup \mathcal{T}_j \text{ with multiple parametrized incoming links,} \\ 0 & \text{otherwise} \end{cases}$$

Merging is represented by an algebraic row- and column operation on \mathfrak{M}

Summary

Identifiability of network model sets is determined by

- Presence and location of external signals,
 - Correlation of disturbances
 - Topology of network: parametrized/fixed modules
-
- Graph-based method for synthesizing allocation of external signals
 - that effectively exploits the presence of fixed (known) modules
 - and can be executed through algebraic operations
 - But reaching the minimum number of excitations is not guaranteed

The end