

Network identifiability - Synthesis

Paul Van den Hof

Doctoral School Lyon, France, 11 April 2024

www.sysdynet.eu
www.pvandenhof.nl
p.m.j.vandenhof@tue.nl



Network identifiability - Synthesis

Consider a network where all nodes are measured.

Question:

Where to allocate external excitation signals in order to guarantee generic identifiability of the network model set?

Graphical approach to cover the network graph with pseudotrees^[1]

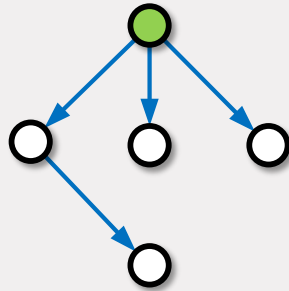
[1] X. Cheng, S. Shi and P.M.J. Van den Hof, IEEE Trans. Automatic Control, Febr 2022.

Network identifiability - Synthesis

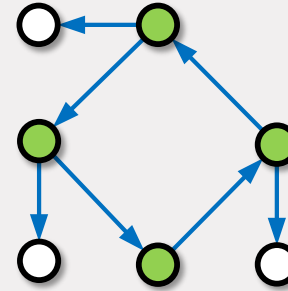
Definition Pseudotree:

A connected simple directed graph with number of vertices ≥ 2 is called a (directed) **pseudotree** if for all vertices i , the number of in-neighbors is ≤ 1 .

Two typical examples:



rooted tree



cycle with outgoing tree

Observation:

An external signal added to any of the roots (green) reaches all vertices in the pseudotree

Network identifiability - Synthesis

Strategy:

- Cover the graph of a network with a set of **disjoint pseudotrees**
- Excite (one of the) root(s) of each pseudotree with an external excitation signal

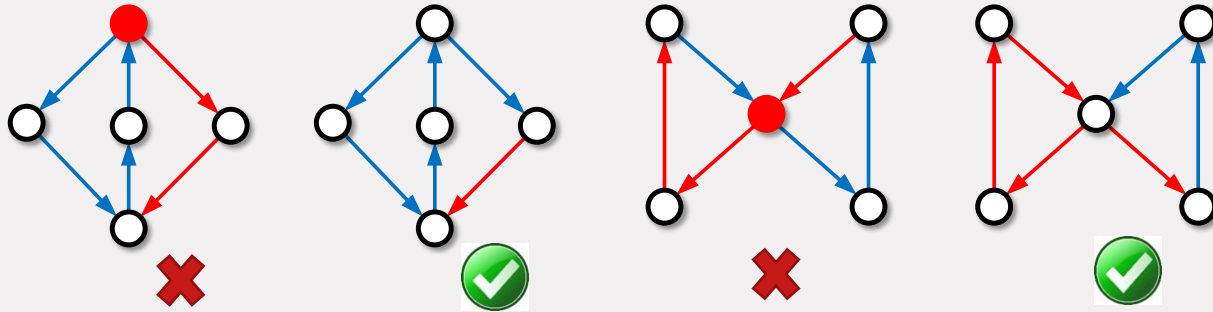
(Edge-) disjoint pseudotrees

Two pseudotrees are (edge-) disjoint if

- They do not share any edges, and
- All outgoing edges of a vertex belong to the same pseudotree

Network identifiability - Synthesis

- Edges are **disjoint** and all out-neighbours of a node are in the same pseudo-tree



- Any network graph can be decomposed into a set of disjoint pseudo-trees

Network identifiability - Synthesis

Synthesis solution for network excitation

A network model set \mathcal{M} is generically identifiable if

- its graph can be covered by K disjoint pseudotrees, and
- there are K independent external signals entering at a root of each pseudotree.

Sketch of Proof:

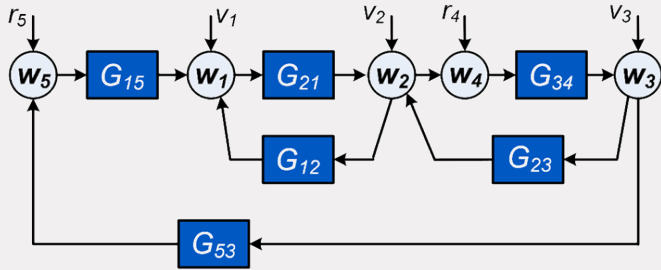
Let $\mathcal{T}_1, \mathcal{T}_2, \dots, \mathcal{T}_K$ be disjoint pseudotrees that cover all the edges of the graph \mathcal{G} and τ_k be an excited root node in pseudotree \mathcal{T}_k .

The definition of disjoint pseudotrees implies that

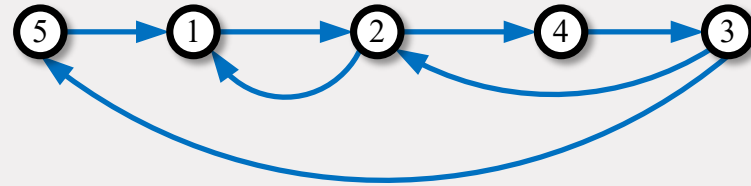
1. two disjoint pseudotrees cannot share common root nodes, i.e., $\tau_i \neq \tau_j$, for all $i \neq j$;
2. the in-neighbors of each node in \mathcal{G} should be in distinct pseudotrees;
3. paths in different disjoint pseudotrees are vertex-disjoint, if they have no common starting or ending nodes.

The above three points guarantees that, for any node j in \mathcal{G} , there exist $|\mathcal{N}_j^-|$ vertex-disjoint paths from the set $\{\tau_1, \tau_2, \dots, \tau_K\}$ to \mathcal{N}_j^- , where \mathcal{N}_j^- is the set of in-neighbors of j . The result holds for all nodes in \mathcal{G} , thus generic identifiability of \mathcal{M} follows.

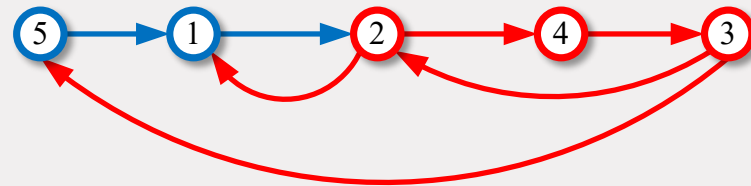
Example: 5 node network (revisited)



When discarding the present external signals, the graph becomes:

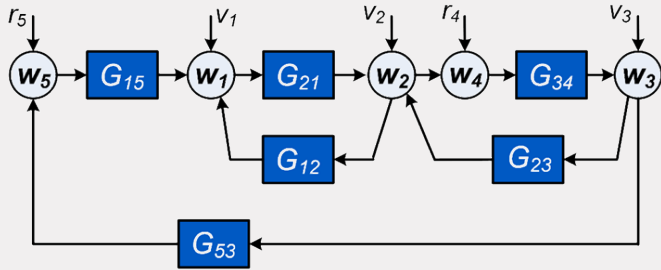


The graph can be covered by
Two disjoint pseudotrees:

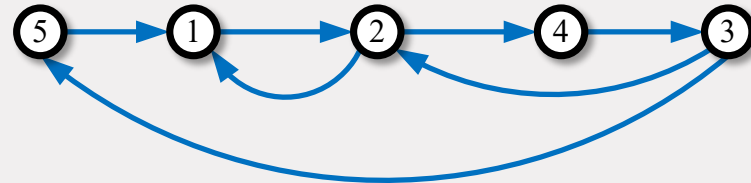


Note: this covering is non-unique!

Example: 5 node network (revisited)

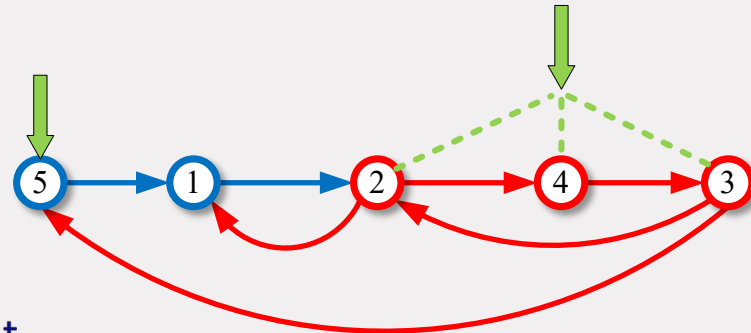


When discarding the present external signals, the graph becomes:

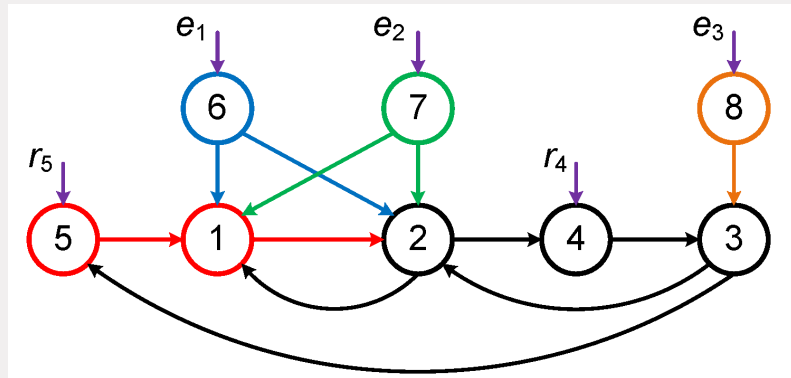


Two independent excitations guarantee network identifiability:

One of $v_2/r_4/v_3$ and r_5 would be sufficient

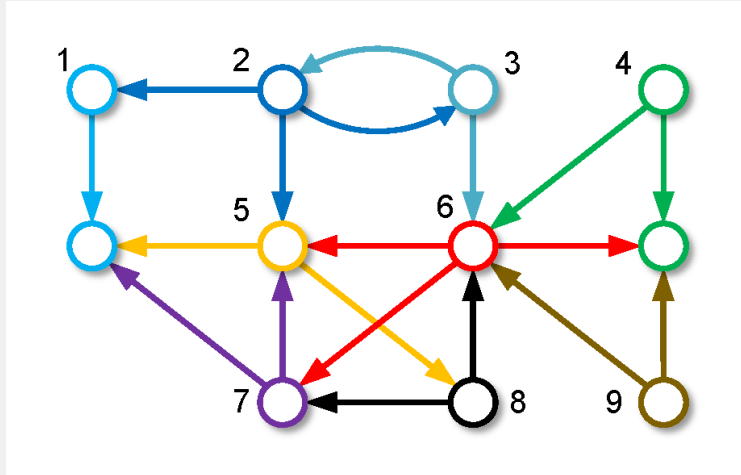


If parametrized noise models are included in the model set, then we use an extended graph, including the white noise disturbance inputs as nodes:

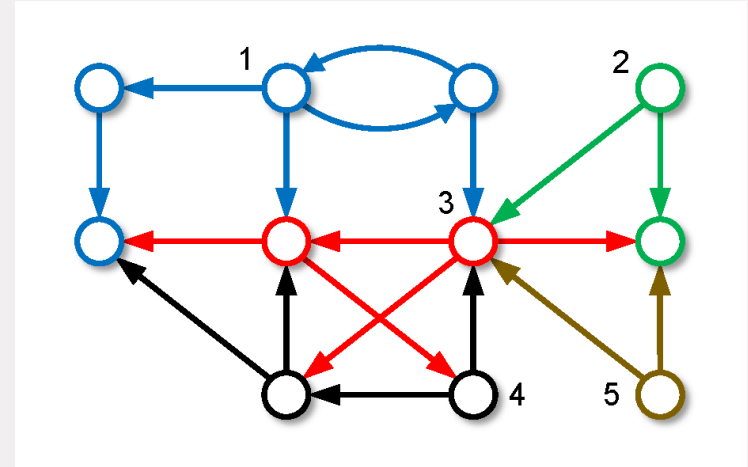


External signals $r_2/r_4/r_3$ and r_5 guarantee generic identifiability

Where to allocate external excitations for network identifiability?



➔
Pseudo-tree
merging
algorithm



Start from an elementary covering
(all outgoing edges from a node in
one pseudotree)

The merging can be done through an automated algorithm

Merging algorithm

Denote a set

$$\mathbb{M} = \{1, 0, \emptyset\}.$$

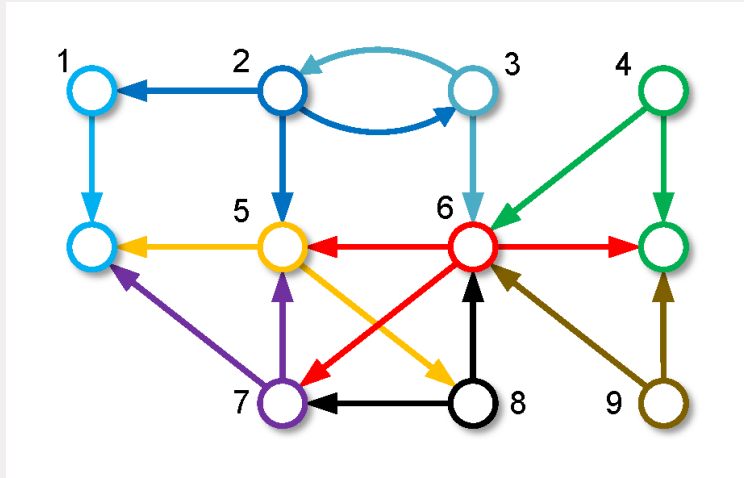
Let $\Pi = \{\mathcal{T}_1, \mathcal{T}_2, \dots, \mathcal{T}_n\}$ be a disjoint pseudotree covering of a directed graph. The **characteristic matrix** of Π is denoted by $\mathcal{M} \in \mathbb{M}^{n \times n}$, whose (i, j) -th entry is defined as

$$\mathcal{M}_{ij} = \begin{cases} 1 & \text{if } \mathcal{T}_i \text{ is mergeable to } \mathcal{T}_j; \\ \emptyset & \text{if } V(\mathcal{T}_j) \cap V(\mathcal{T}_i) = \emptyset; \\ 0 & \text{otherwise.} \end{cases}$$

and define a commutative operator on \mathbb{M} according to $1 \odot 1 = 1, 1 \odot 0 = 0, 1 \odot \emptyset = 1,$
 $0 \odot 0 = 0, \emptyset \odot 0 = 0, \emptyset \odot \emptyset = \emptyset.$

defining a componentwise multiplication operation on rows of \mathcal{M}

Start of the algorithm: elementary covering



$$M = \begin{bmatrix} 0 & 1 & \emptyset & \emptyset & 0 & \emptyset & 0 & \emptyset & \emptyset \\ 0 & 0 & 1 & \emptyset & 0 & 0 & 0 & \emptyset & \emptyset \\ \emptyset & 1 & 0 & 0 & \emptyset & 0 & \emptyset & 0 & 0 \\ \emptyset & \emptyset & 0 & 0 & \emptyset & 0 & \emptyset & 0 & 0 \\ 0 & 1 & \emptyset & \emptyset & 0 & 1 & 0 & 0 & \emptyset \\ \emptyset & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & \emptyset & \emptyset & 0 & 0 & 0 & 1 & \emptyset \\ \emptyset & \emptyset & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ \emptyset & \emptyset & 0 & 0 & \emptyset & 0 & \emptyset & 0 & 0 \end{bmatrix}.$$

Merging algorithm

Merging of the i -th pseudotree into the j -th one now comes down to

- Replace row j by $M_{i*} \odot M_{j*}$
- Replace column j by $M_{*i} \odot M_{*j}$
- Remove the i -th row and column of M

Ordering of the merging:

- Select the row with a (single) 1 entry and a maximum number \emptyset entries, and merge this row;

At the end, the matrix M will have no more 1 entries.

Start of the algorithm: elementary covering

Given a graph \mathcal{G} with the adjacency matrix $A(\mathcal{G})$. Denote

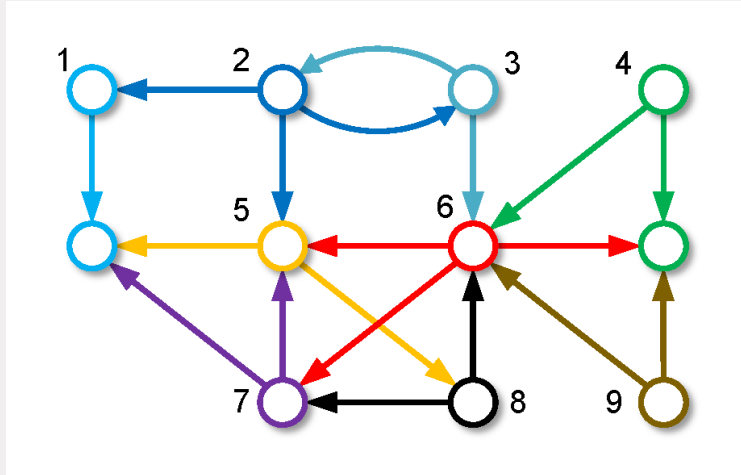
$$a_{ij} = ([A(\mathcal{G}) + I\mathbf{i}]_{\star i})^\top [A(\mathcal{G}) + I\mathbf{i}]_{\star j},$$

The characteristic matrix \mathcal{M} is formulated as follows:

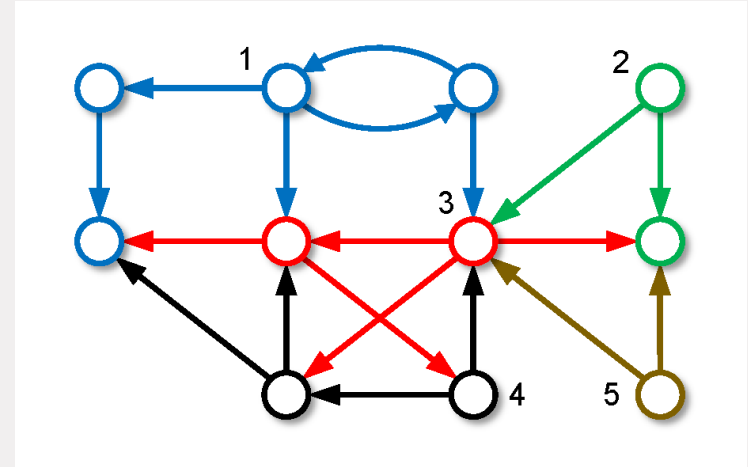
$\mathcal{M}_{ii} = 0$ for all i , while for $j \neq i$:

$$\mathcal{M}_{ij} = \begin{cases} 1, & \text{Re}(a_{ij}) = 0, \text{ and } \text{Im}(a_{ij}) \neq 0, \text{ and } [A(\mathcal{G})]_{ij} \neq 0. \\ 0, & \text{Re}(a_{ij}) \neq 0 \text{ or } \{\text{Re}(a_{ij}) = 0, \text{ and } \text{Im}(a_{ij}) \neq 0, \text{ and } [A(\mathcal{G})]_{ij} = 0\}. \\ \emptyset, & a_{ij} = 0, \end{cases}$$

Where to allocate external excitations for network identifiability?



➔
Pseudo-tree
merging
algorithm

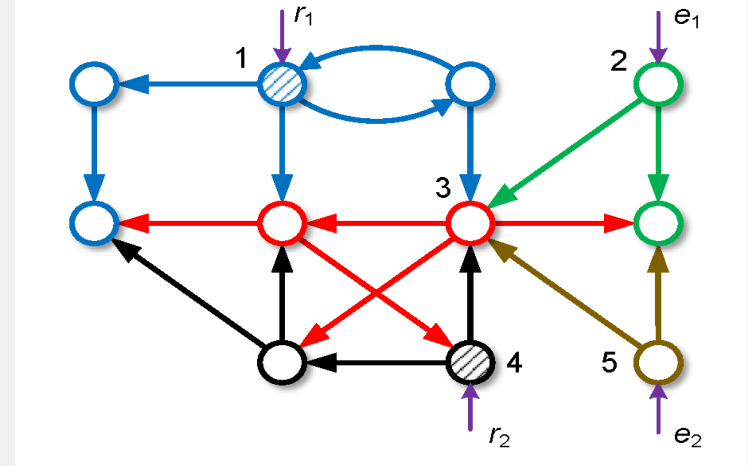


If white noises e_2 and e_5 are present, then it suffices to excite r_1 , r_3 and r_4 .

Where to allocate external excitations for network identifiability?

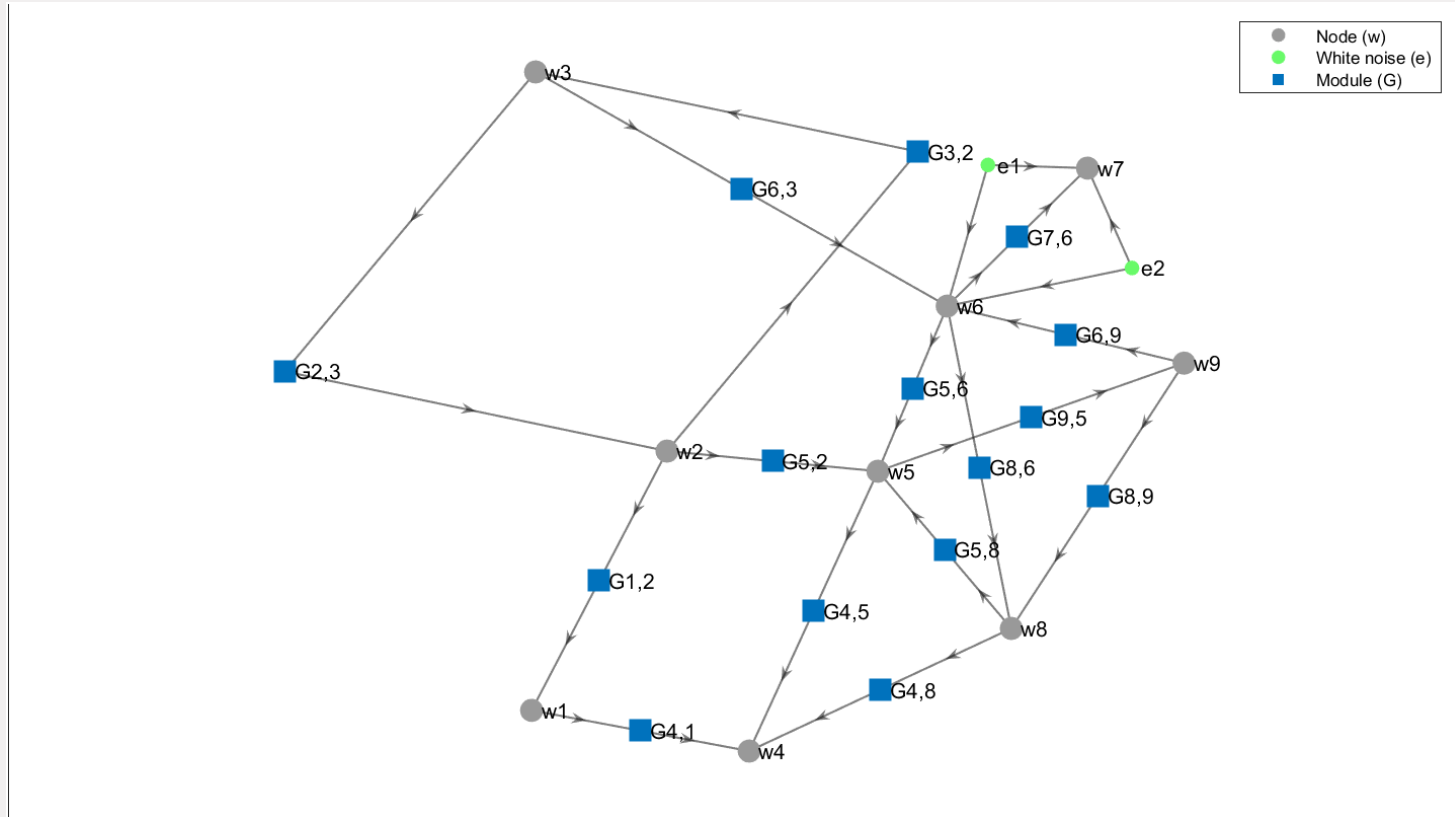
After selecting the roots of the pseudotrees:

Verify whether all root excitations are necessary for satisfying the path-based identifiability condition (# vertex disjoint paths)



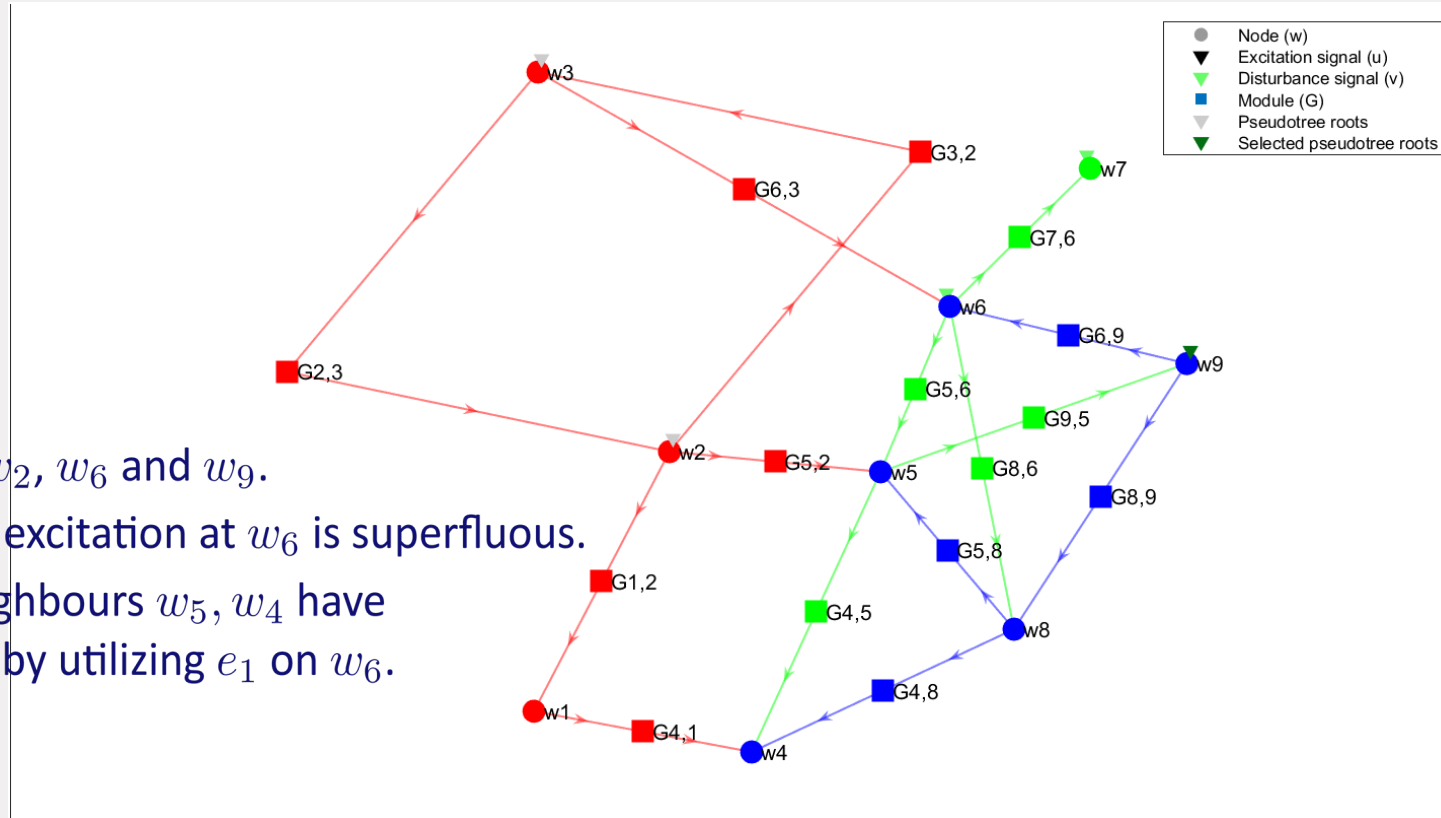
Since the path-based condition is satisfied for all nodes in pseudotree 3, even without the presence of r_3 , this excitation can be removed.

Algorithm example



Algorithm example

Excitation at w_3 or w_2, w_6 and w_9 .
Analysis shows that excitation at w_6 is superfluous.
Nodes with 3 in-neighbours w_5, w_4 have sufficient excitation by utilizing e_1 on w_6 .



Summary identifiability synthesis algorithm

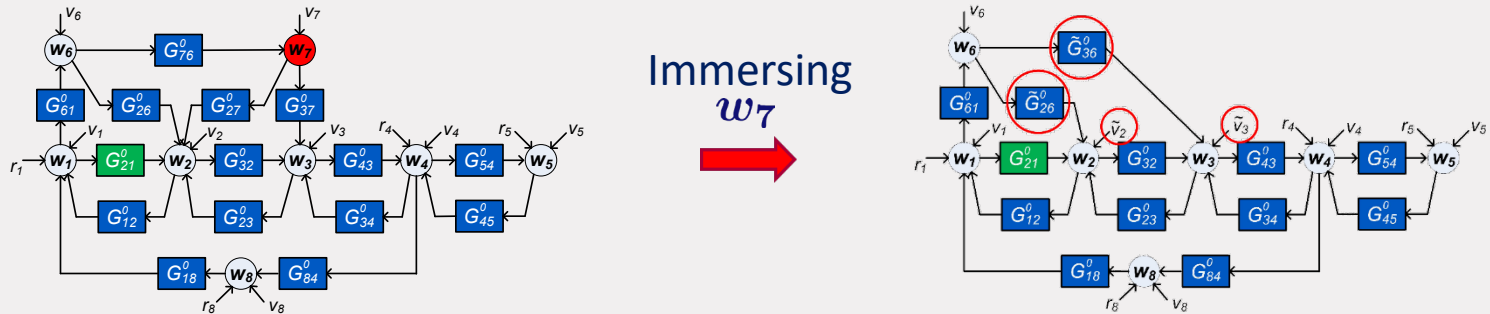
- Attractive graphical approach for verifying generic identifiability conditions.
- As well as for synthesizing the required experimental setup (allocating external signals), starting from existing disturbances.
- The results also apply to the situation of non-parametrized / fixed modules in \mathcal{M} ; The fixed modules can then be excluded from the graph-covering.
- A less conservative way of including fixed modules is available by extending the concept of a pseudotree, to a graph with at most one parametrized link from an in-neighbor^[1]; this is implemented in the toolbox.

[1] Dreef et al., L-CSS, 2022.

Discussion identifiability

If node signals can not all be measured? (partial node measurement)

- Situation can be treated as separate problem^[1], leading to statements that for identifiability each node should be measured or excited.
- Situation can partly be analysed by using the concept of **immersion**, i.e. removing a non-measured node from the network while keeping the other node signals invariant.^[2]



[1] Bazanella et al., CDC 2019; Mapurunga et al., IFAC POL, Jan 2021; L-CSS, 2022; Cheng et al., IEEE-TAC, under review, 2022.

[2] Dankers et al., TAC 2016