



POTSDAM INSTITUTE FOR
CLIMATE IMPACT RESEARCH

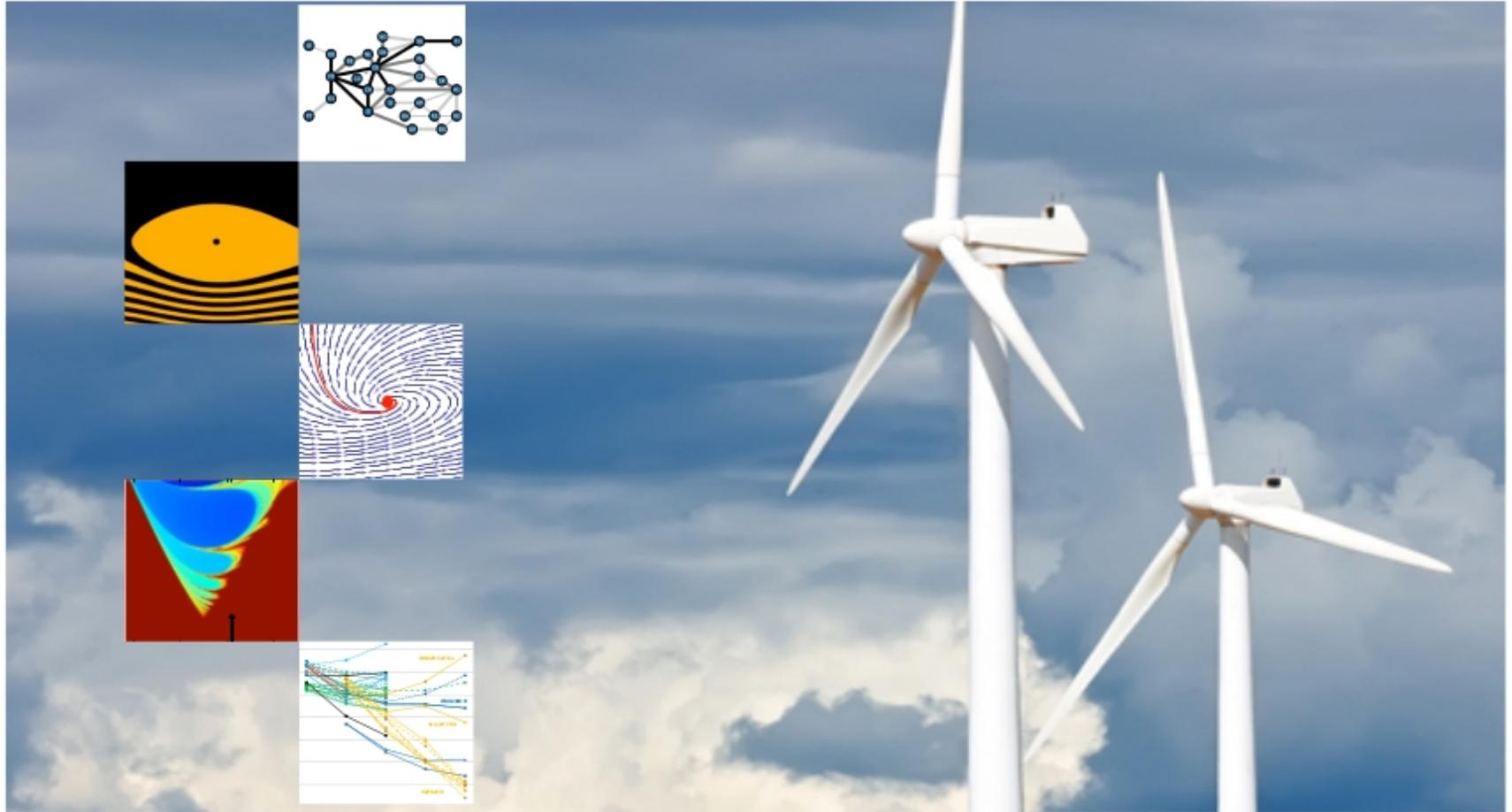
BASIN STABILITY: LARGE PERTURBATIONS IN POWER GRIDS

Paul Schultz

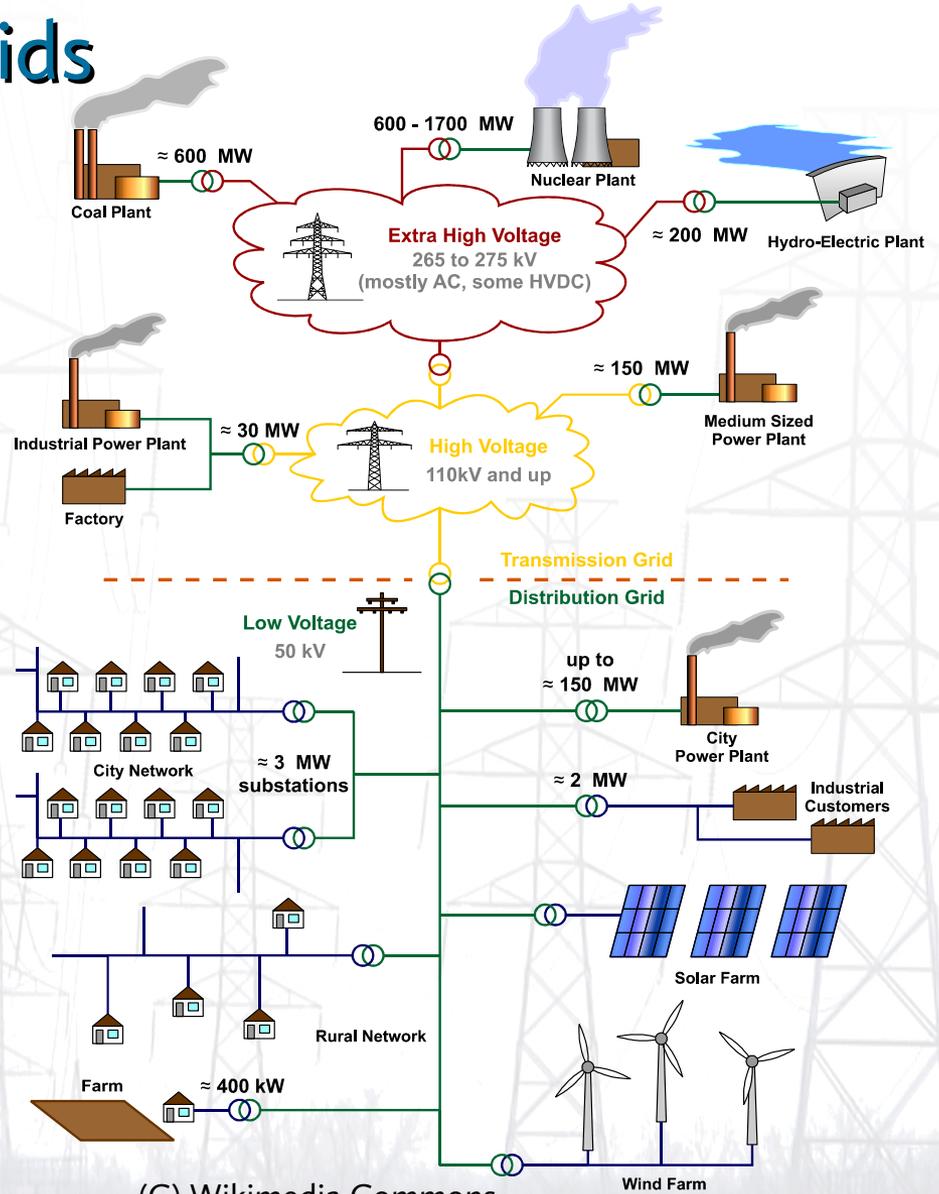
PIK Research Domain IV, Transdisciplinary Concepts & Methods

April 21, 2015

Collective Nonlinear Dynamics of Power Grid Networks: Stability, Efficiency and Risks



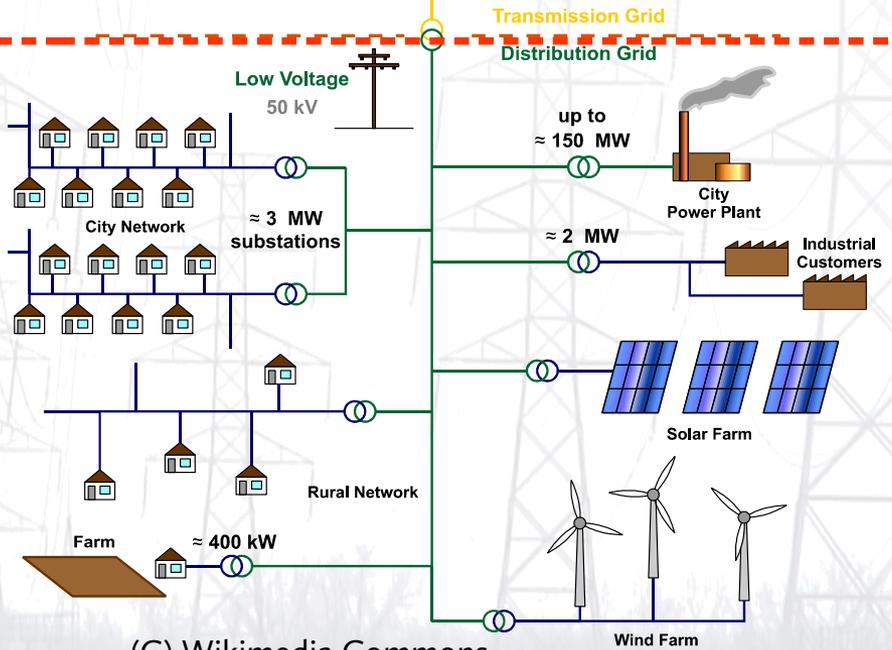
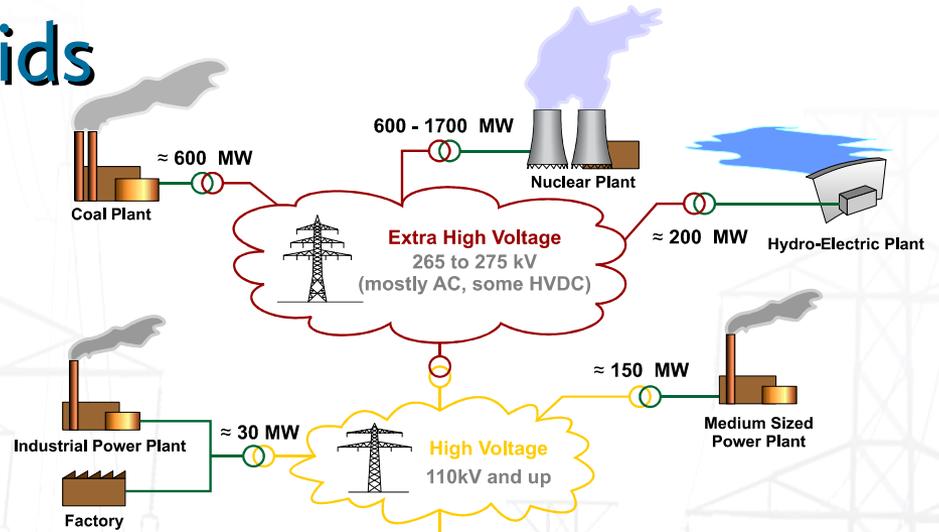
Structure of Power Grids



(C) Wikimedia Commons



Structure of Power Grids



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A real world network



1. How to model Power Grids?

Power Grid Dynamics

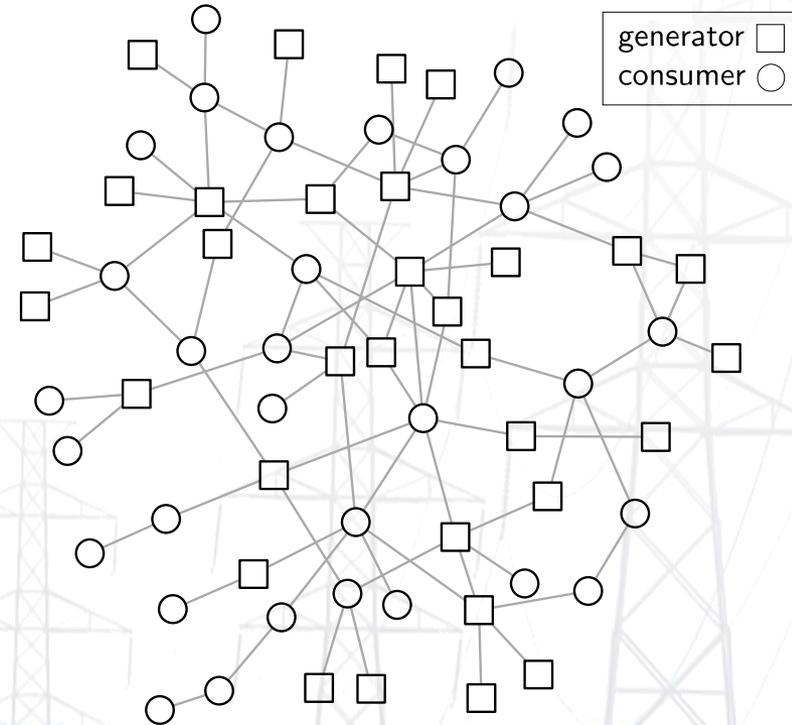
$$\ddot{\phi}_i = P_i - \alpha_i \dot{\phi}_i - \sum_{j=1}^N K_{ij} \sin(\phi_i - \phi_j + \phi_{ij}) \quad \phi_{ij} := \arctan \frac{R_{ij}}{X_{ij}}$$

- physical model of generator dynamics
- effective network model
- simplifications:
 - constant couplings
 - dynamics close to Ω_{sync}

synchronised state:

$$\forall i : \dot{\phi}_i = 0$$

$$\forall (i, j) : \phi_i - \phi_j = \text{const.} \neq 0$$



Ref.: Filatrella et al., *EPJ B* 61(4), 2008

Ref.: Machowski et al, Wiley, New York, 2008

Ref.: Nishikawa & Motter, *NJP* (17) 15012, 2015

Power Grid Dynamics

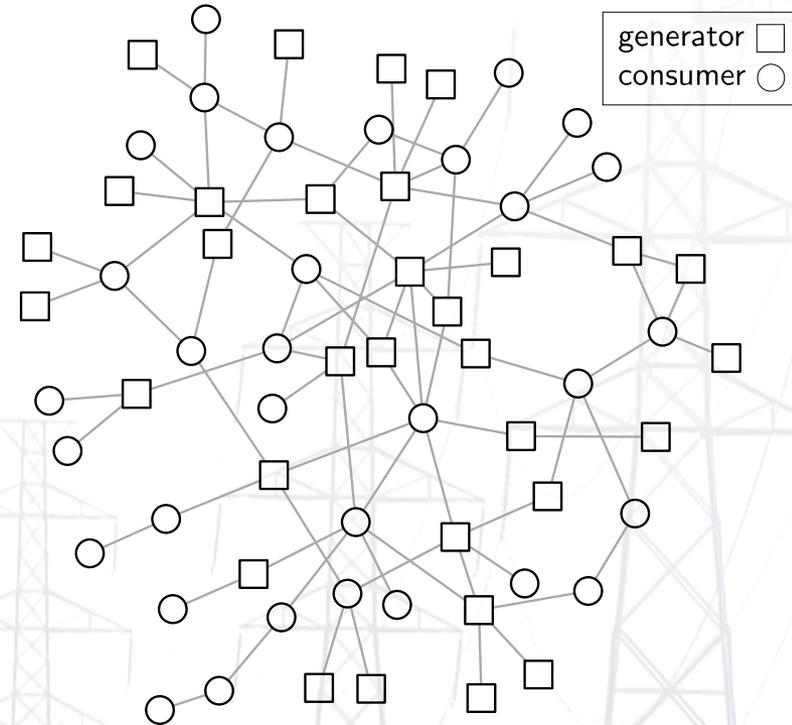
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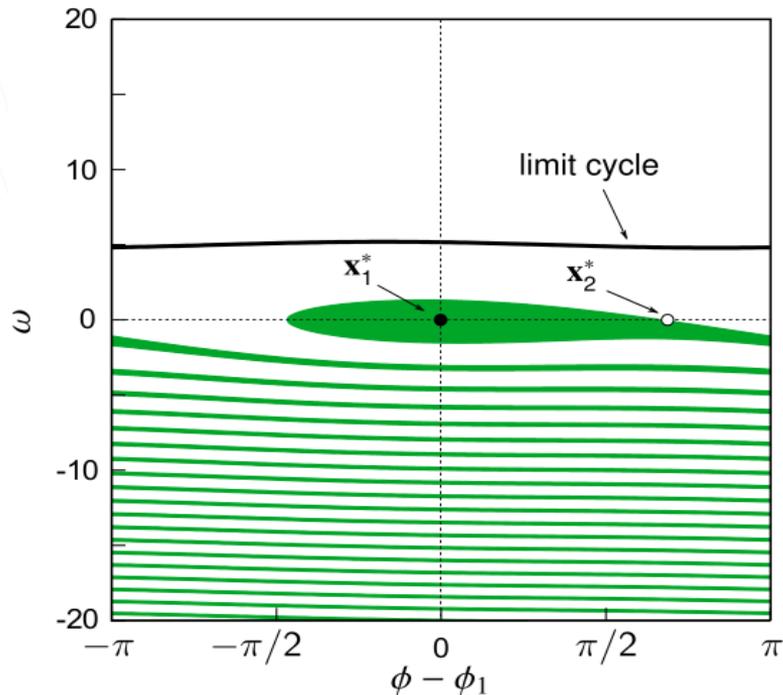


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Estimation of Basin Stability

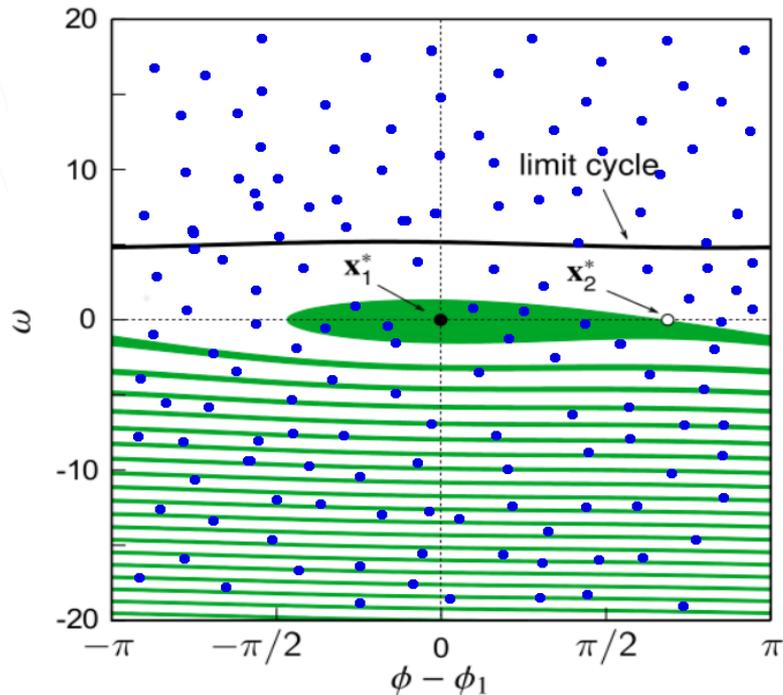


bistable system:

final state depends on IC

$$\mathbf{B} = \{x \in \mathcal{S} | \Phi_t(x) \rightarrow \mathbf{I}\}$$

Estimation of Basin Stability



bistable system:

final state depends on IC

$$\mathbf{B} = \{x \in S \mid \Phi_t(x) \rightarrow \mathbf{I}\}$$

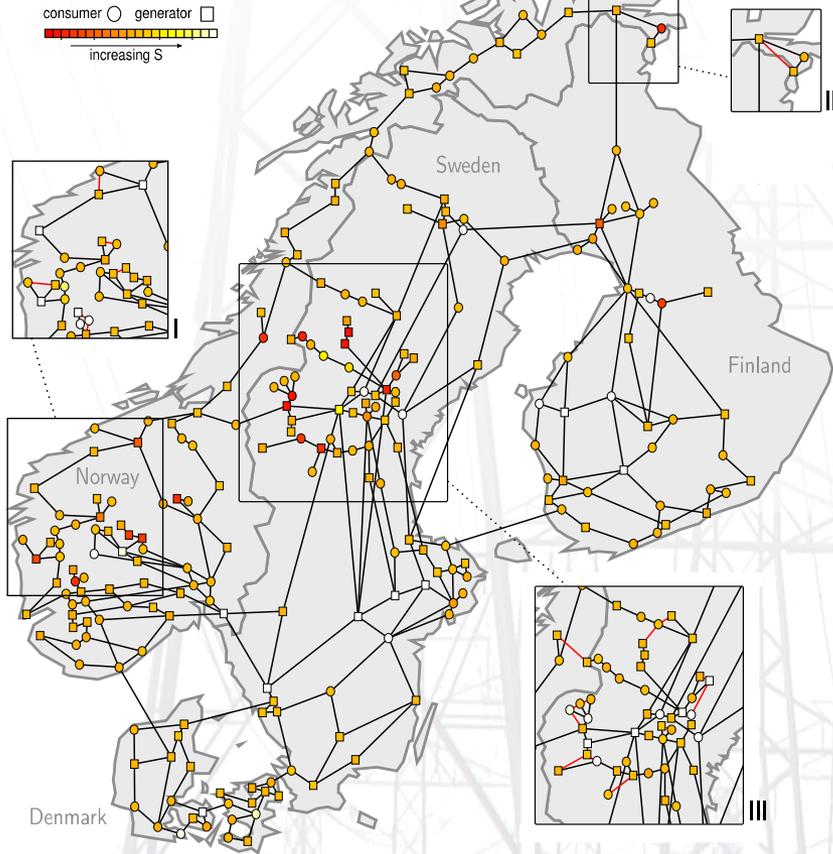
$$BS = \frac{\# \text{blue in green}}{\# \text{all blue}}$$

$$S_{\mathbf{B}} = \int_S \chi_{\mathbf{B}} \rho(x) dx$$

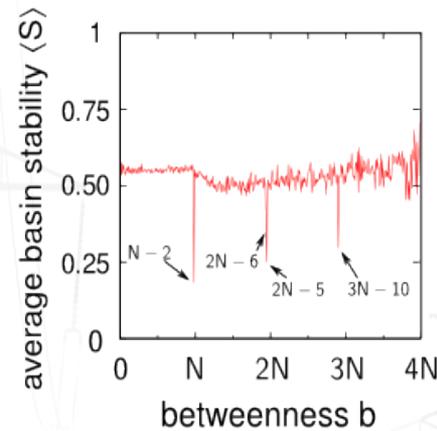
single node basin stability

$$\mathbf{B}_i := \{(\phi_i, \omega_i) \in \mathbf{B} \mid \forall_{j \neq i} (\phi_j, \omega_j) = (\phi_j^*, 0)\}$$

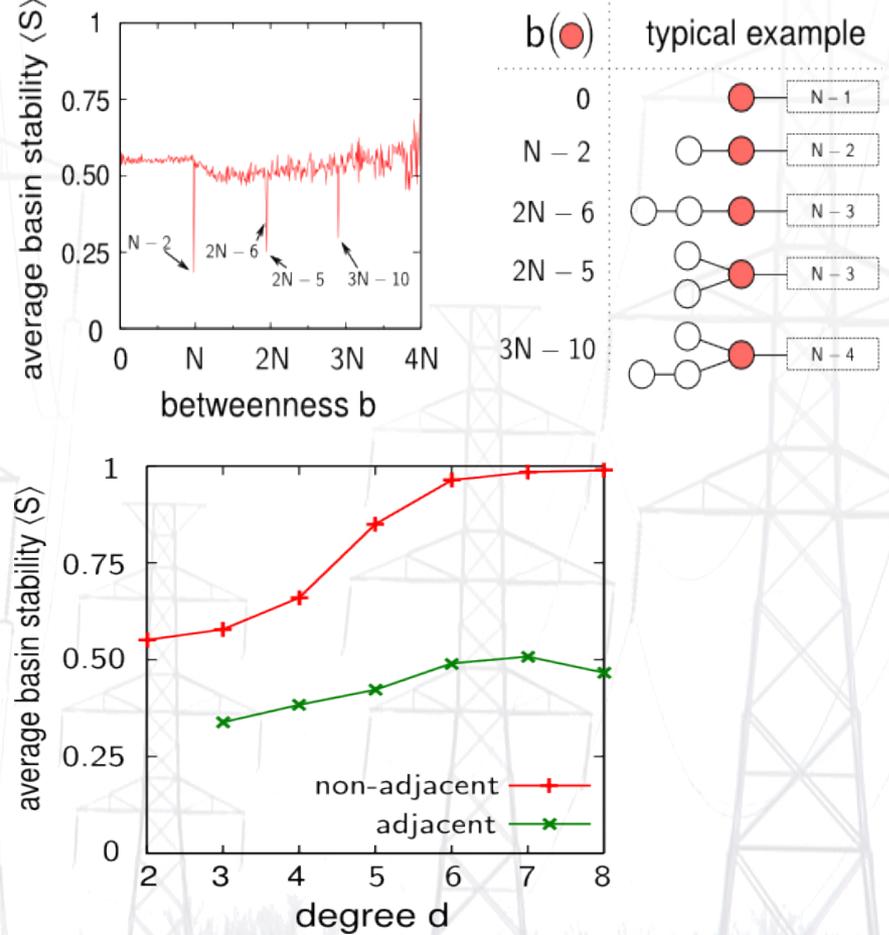
Avoid Dead Ends



A



B



Ref.: Menck et al., Nature Com. (5), 2014

2. Random Network Model

Random Network Model: Key Ideas

- Model wide range of observed network properties
→ degree distribution, sparsity, connectivity, aspl, ...
- Create realistically appearing power grids with spatial embedding
→ supports random as well as given node locations
- Low computational complexity allows for extensive simulations
- Plausible construction mechanism using only few assumptions
- Two stages: initialisation and growth
→ tunable trade-off: cost minimisation vs. redundancy

Ref.: Rosas-Casals, Topological Complexity of the Electricity Transmission Network., UPC (2009)

Model Implementation

- Initialisation with **minimum spanning trees**
→ minimise overall edge weight (i.e. length)

Ref.: Borůvka 1926, Kruskal 1956, Prim 1957

- Transmission lines might be **split** if a new power plant appears closeby
- Trade-off between **redundancy** and **costs**

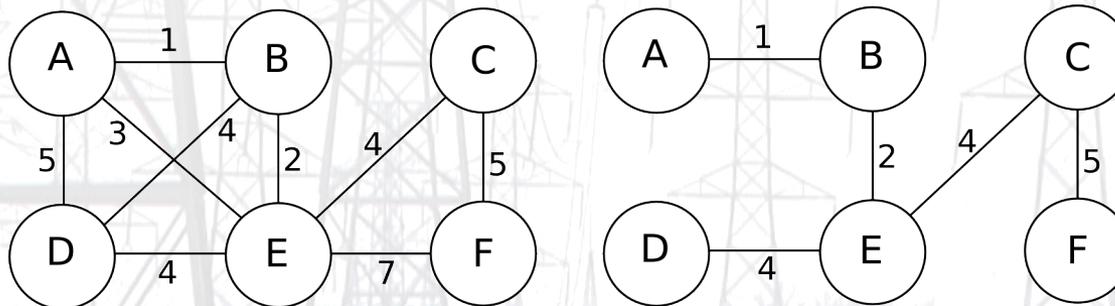
$$\text{maximise } f = \frac{(\text{new redundant lines})^r}{\text{spatial distance}}$$

Minimum Spanning Tree MST

- Invented to design Moravia (Mähren) Power Grid

Ref.: Borůvka (1926), Kruskal (1956), Prim (1957)

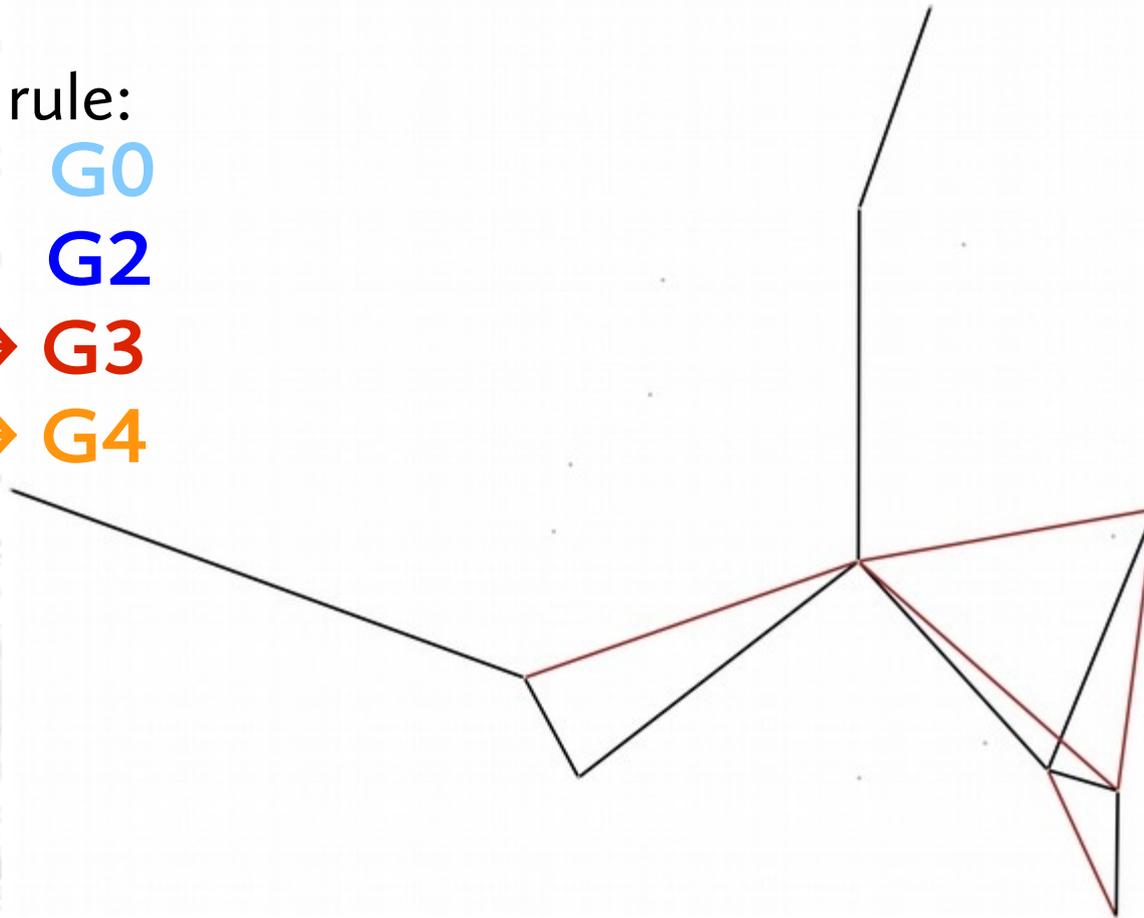
- Draws connected graph (i.e. a tree) with $N-1$ edges between a set of nodes that minimises a given edge weight \rightarrow spatial distance, resistance, general cost function
- Unique if all edge weights are mutually distinct



Growth Algorithm

attachment rule:

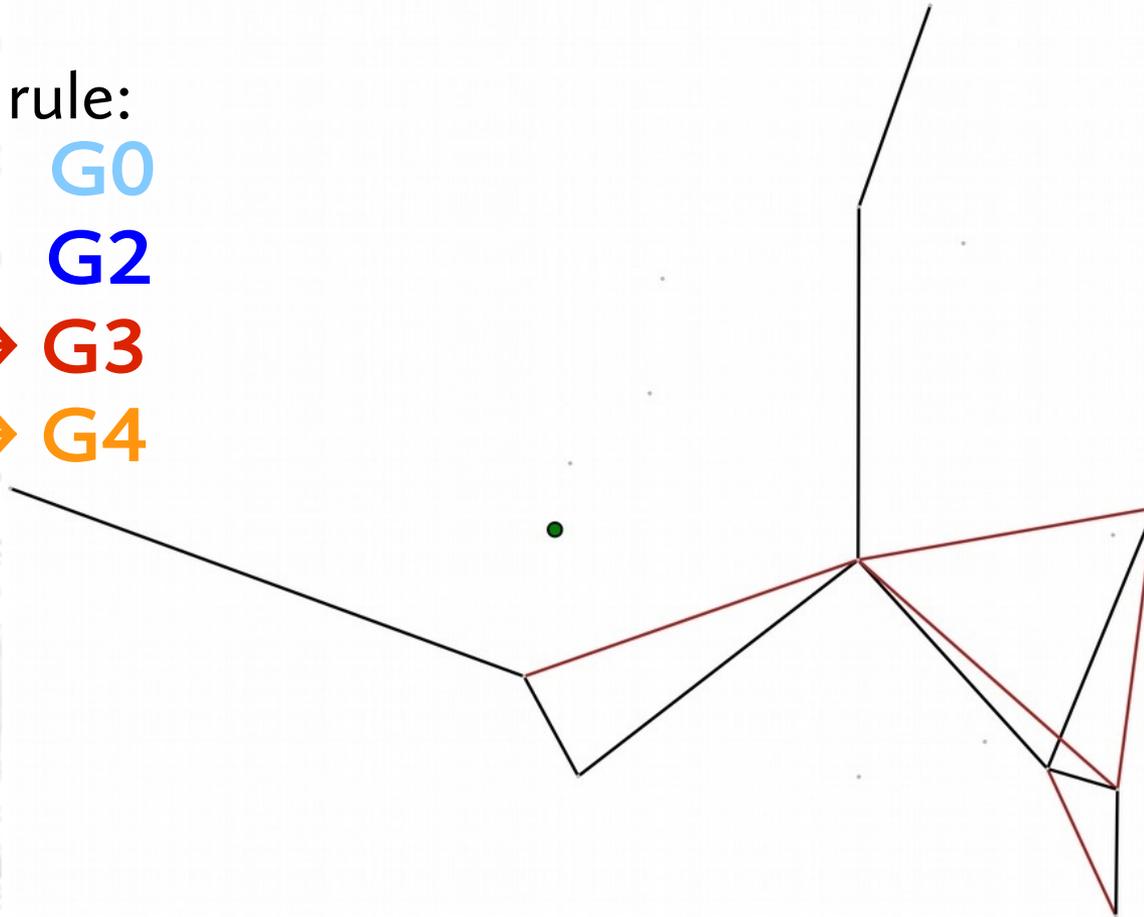
$s \rightarrow G0$
 $1-s \rightarrow \left\{ \begin{array}{l} G2 \\ p \rightarrow G3 \\ q \rightarrow G4 \end{array} \right.$



Growth Algorithm

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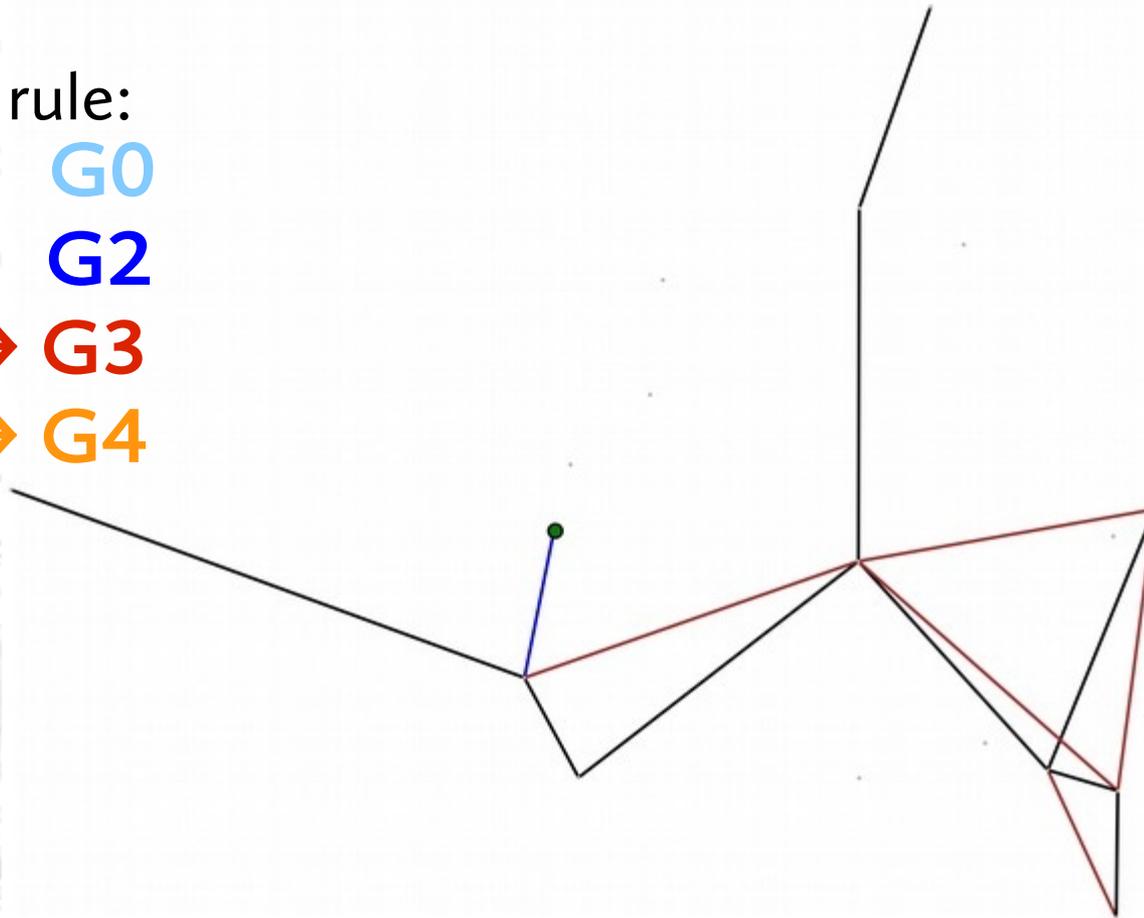
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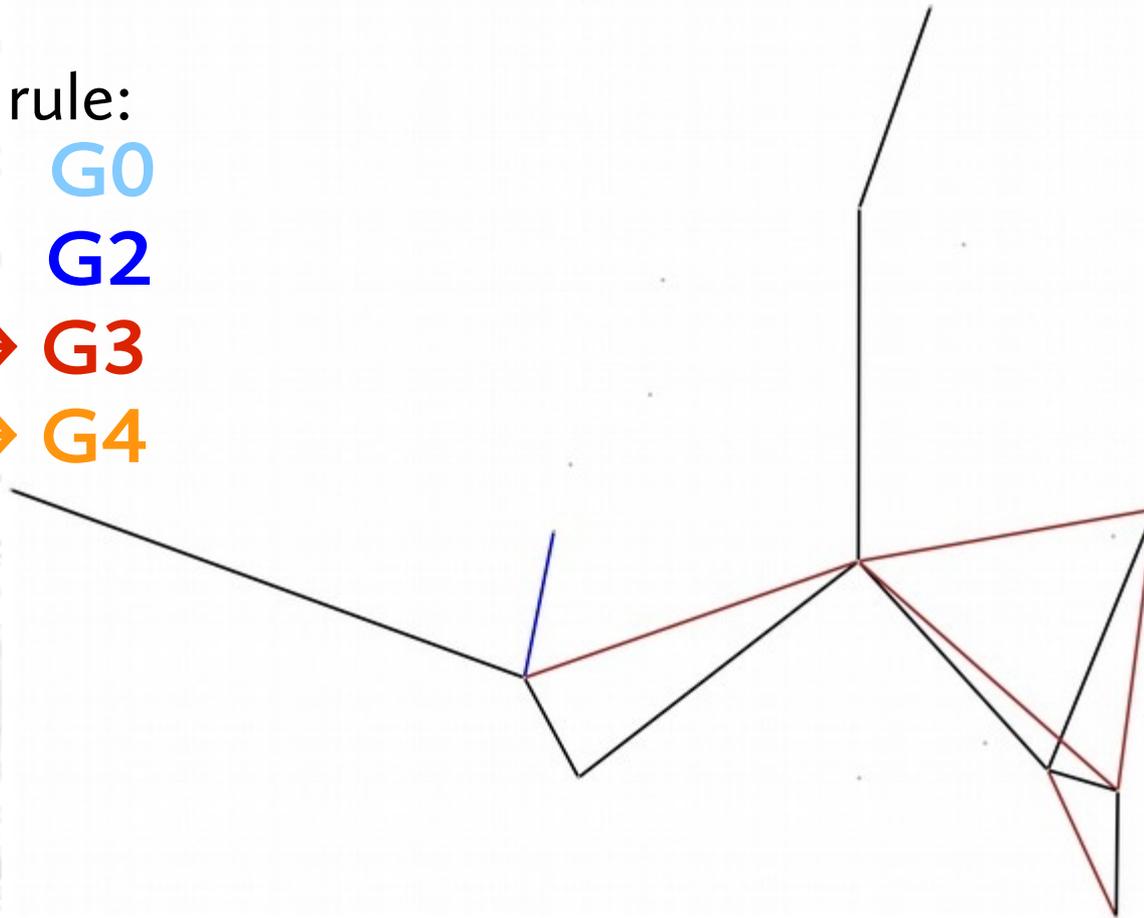
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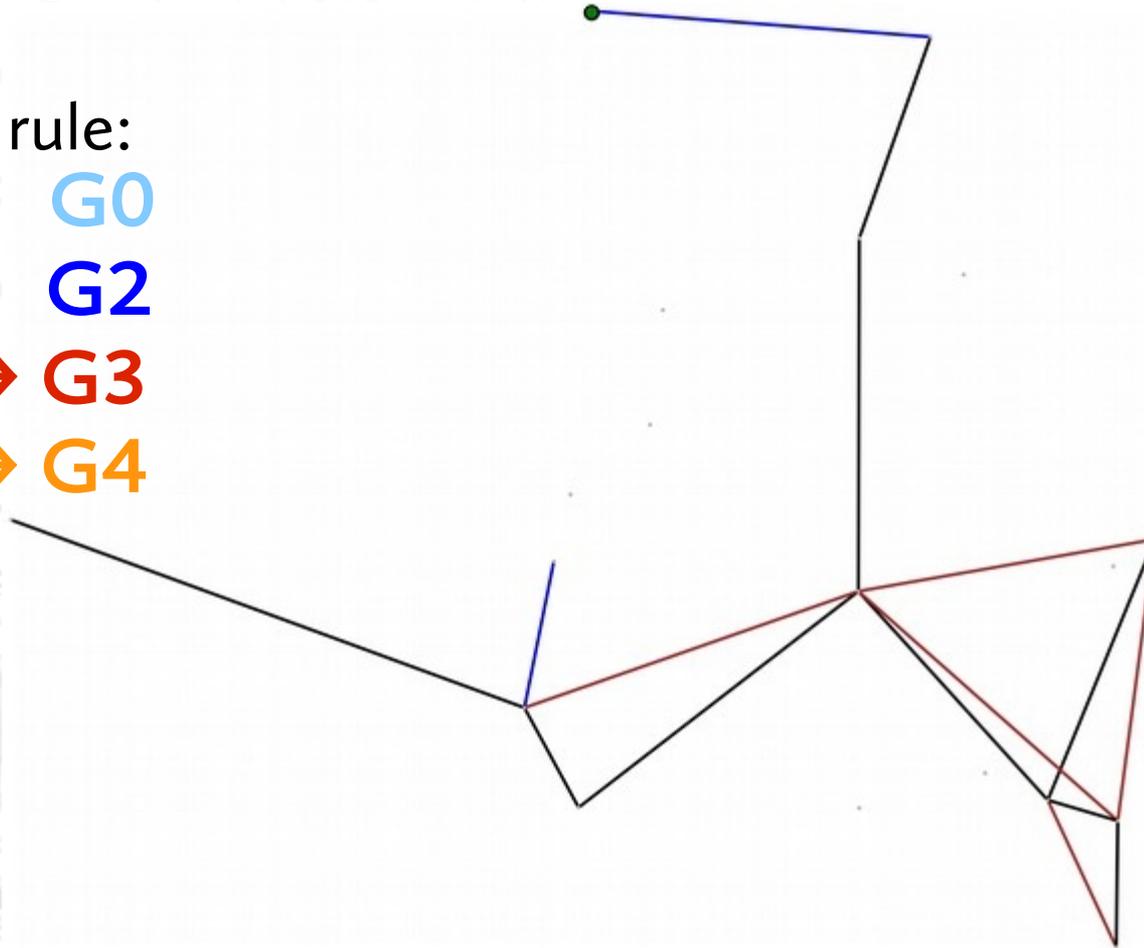
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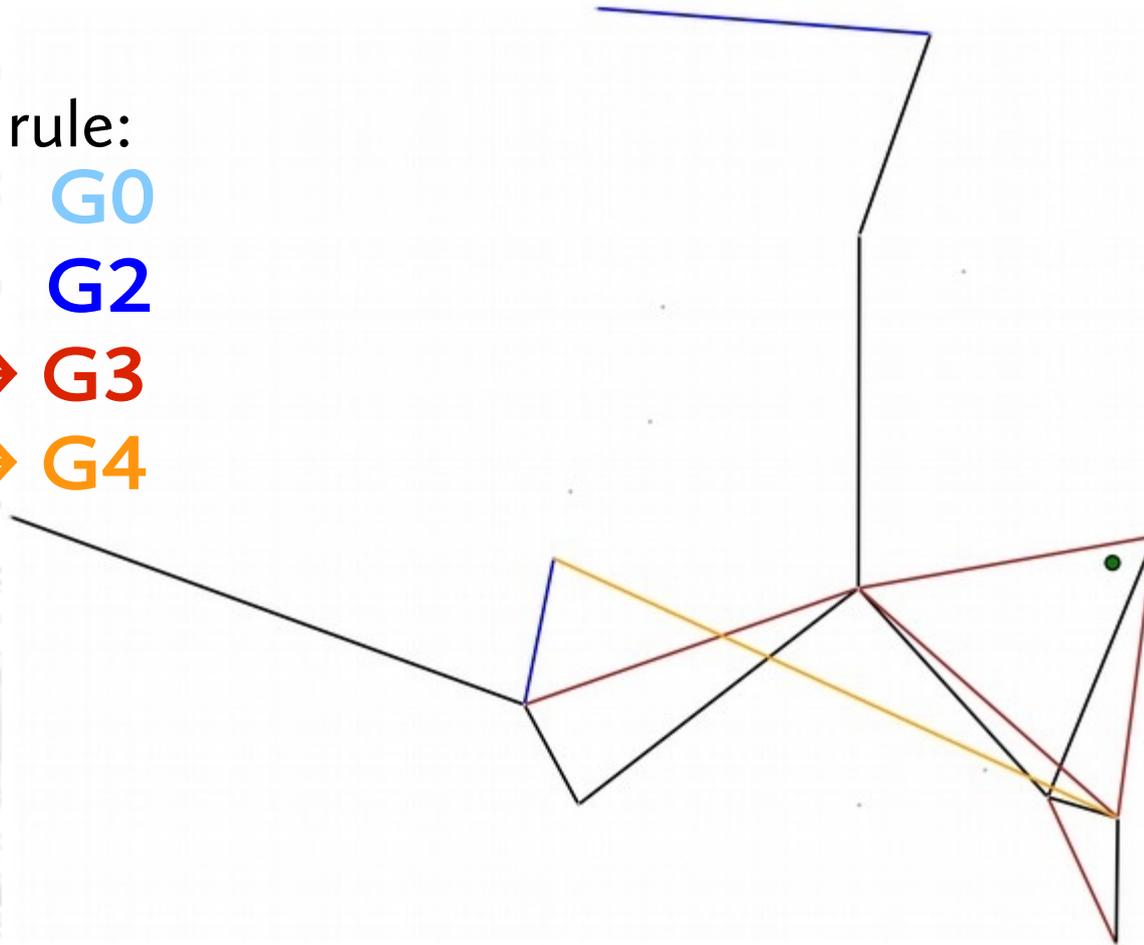
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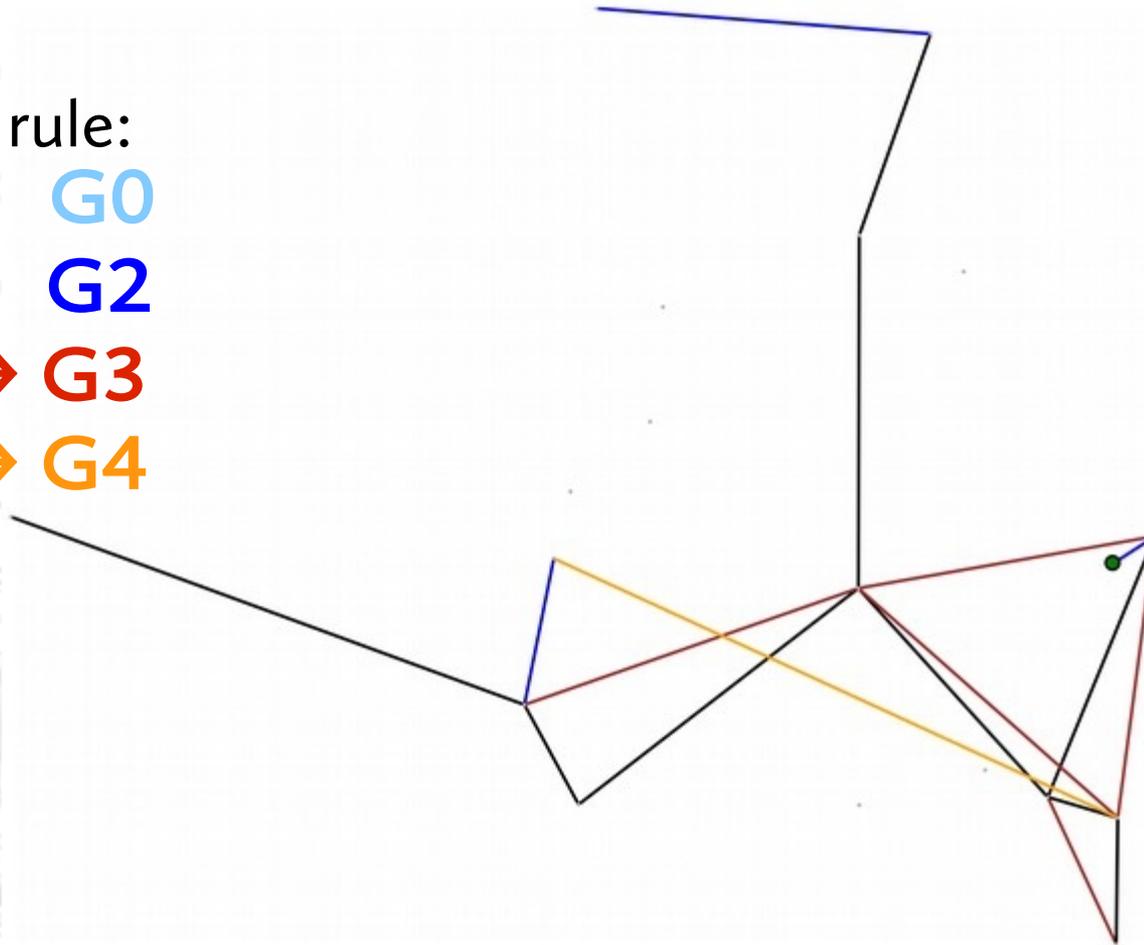
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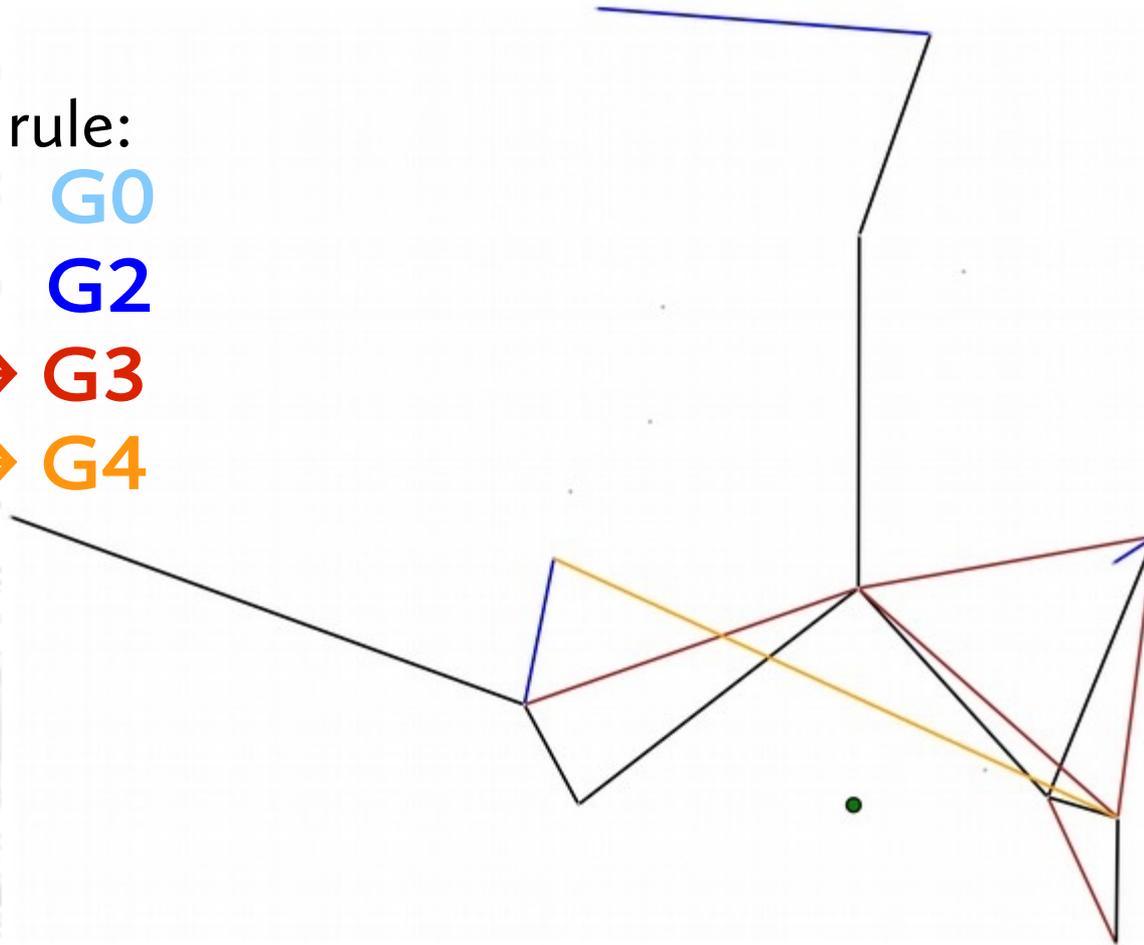
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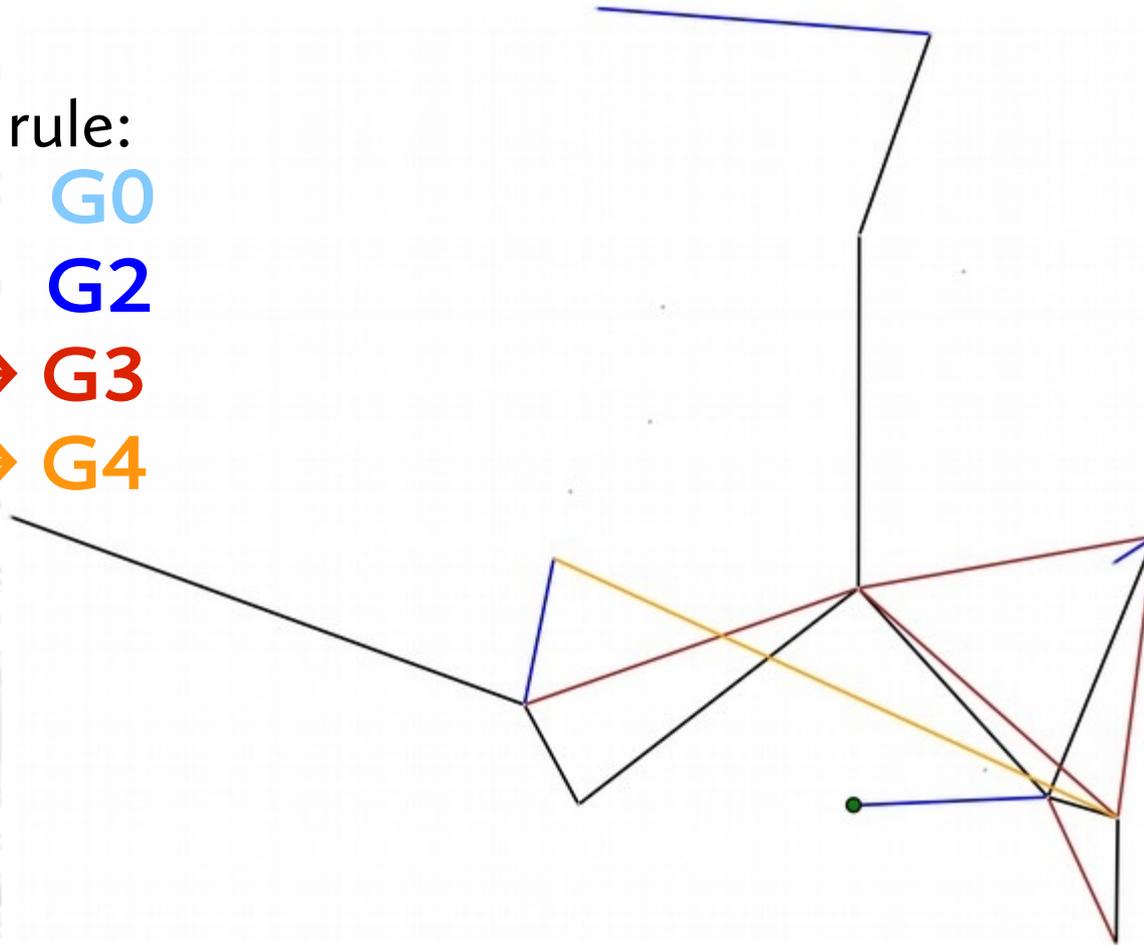
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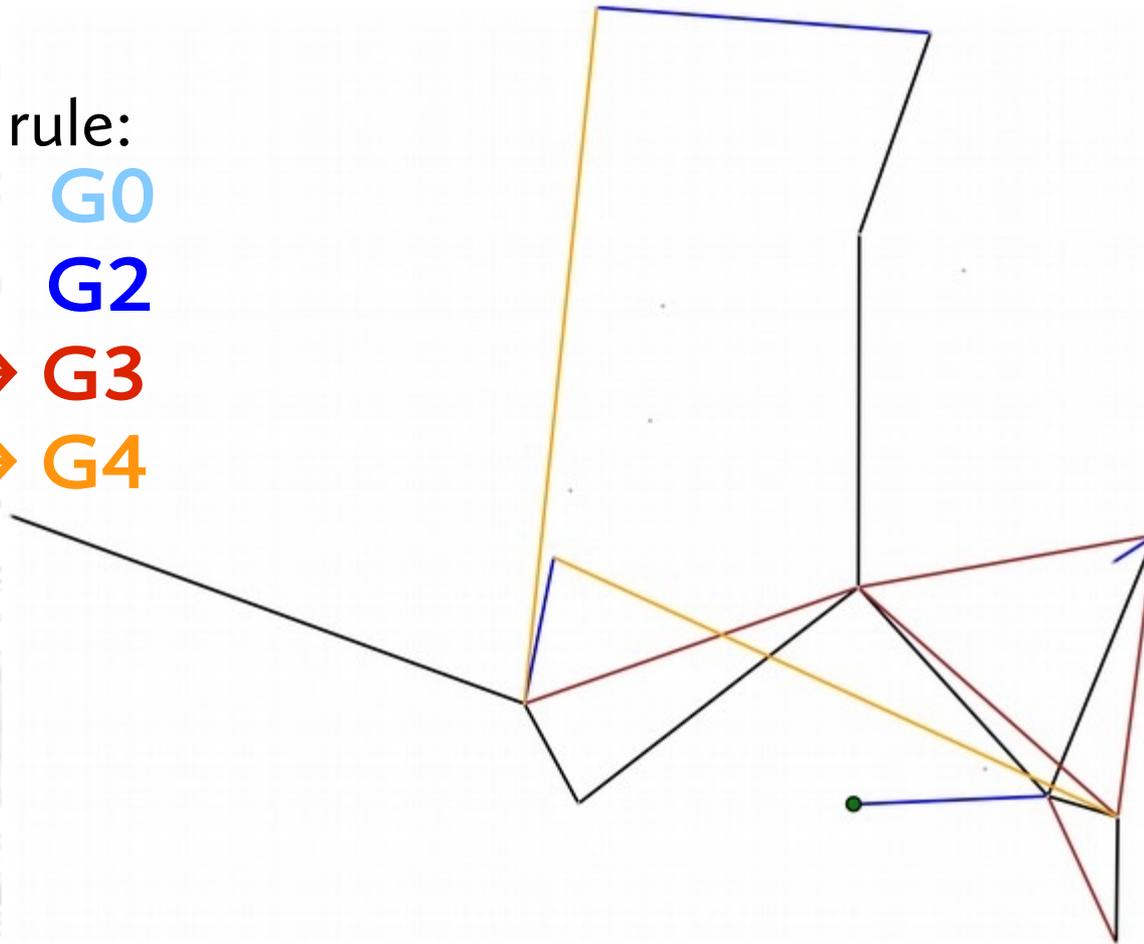
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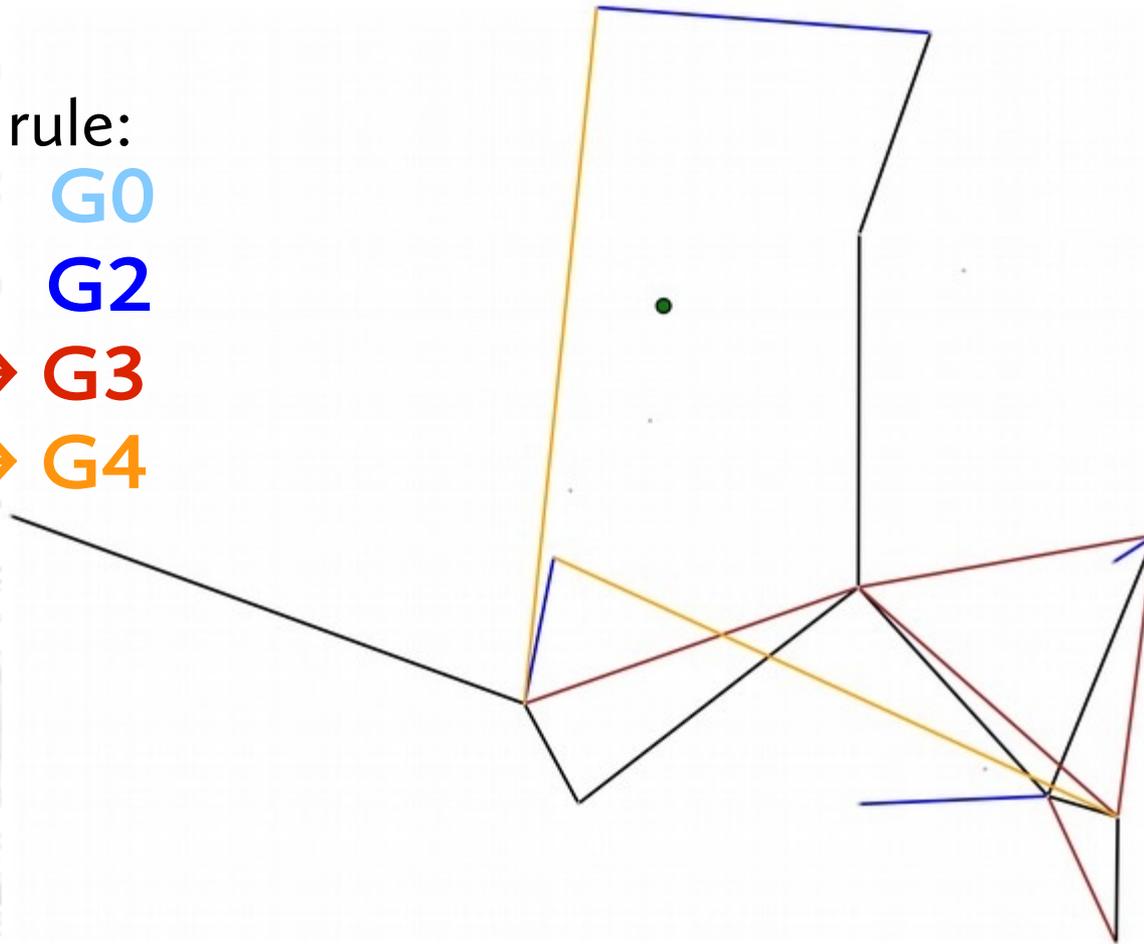
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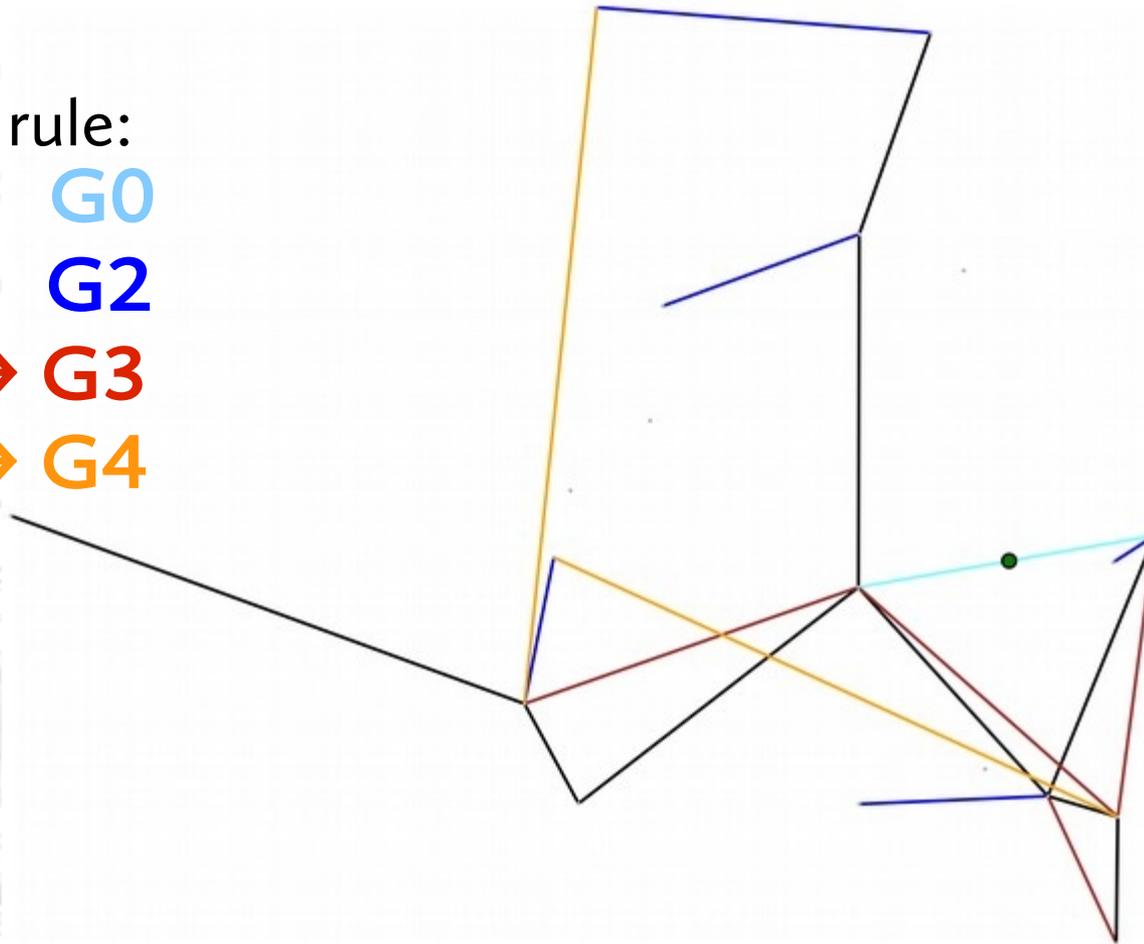
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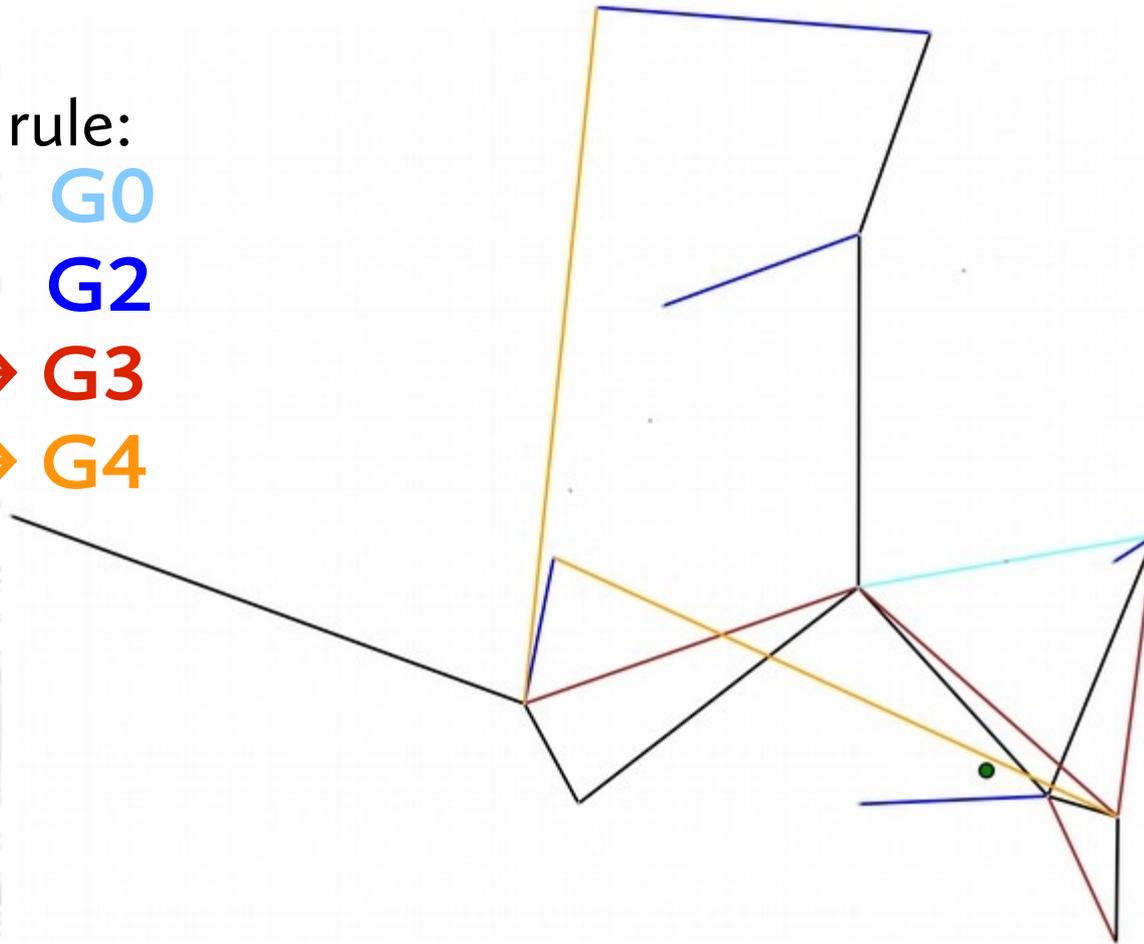
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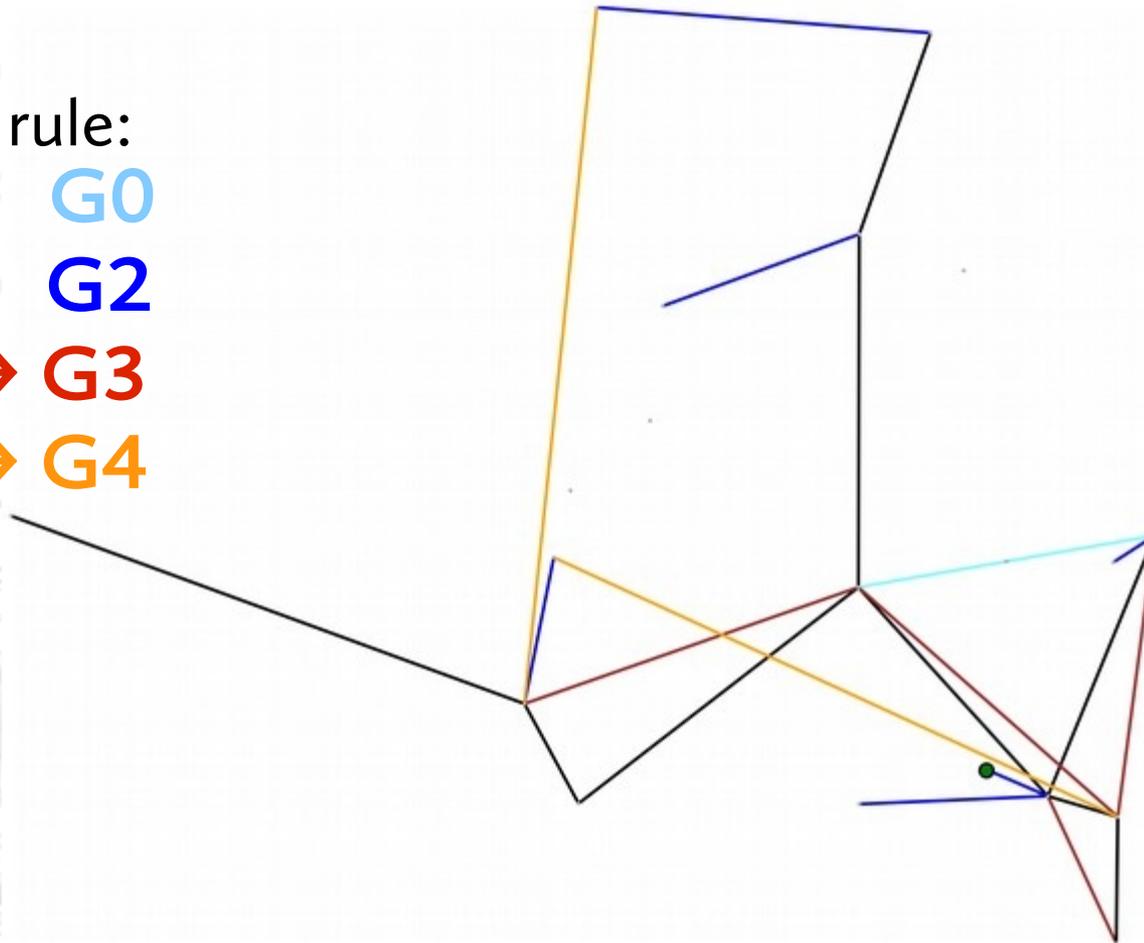
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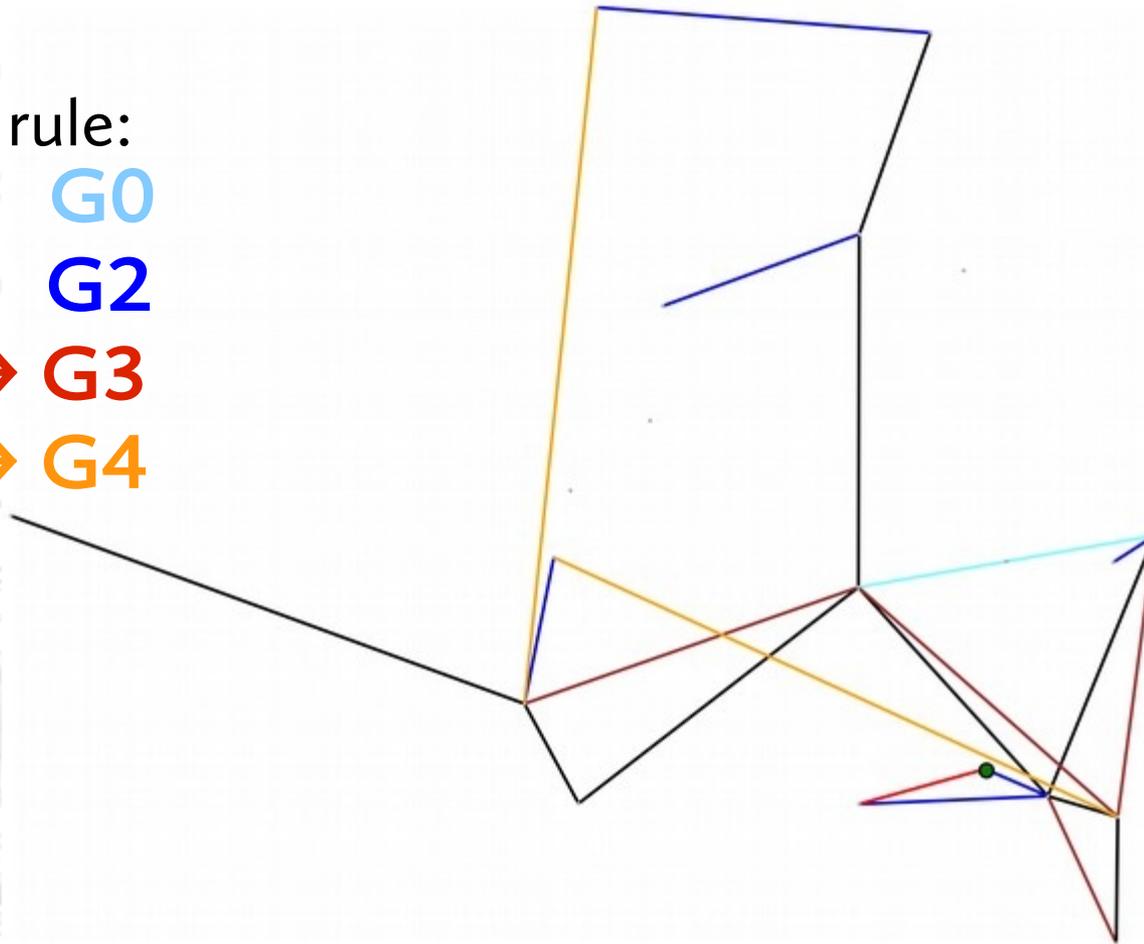
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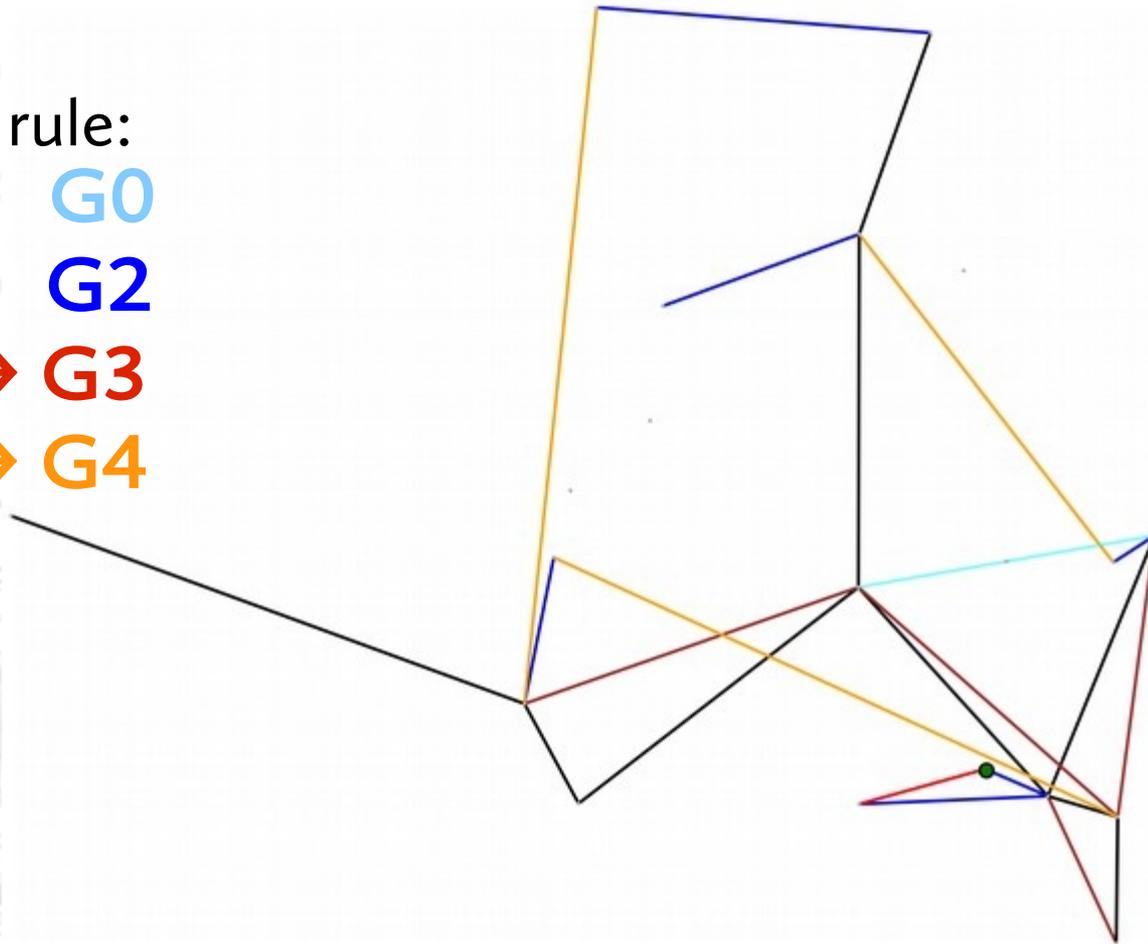
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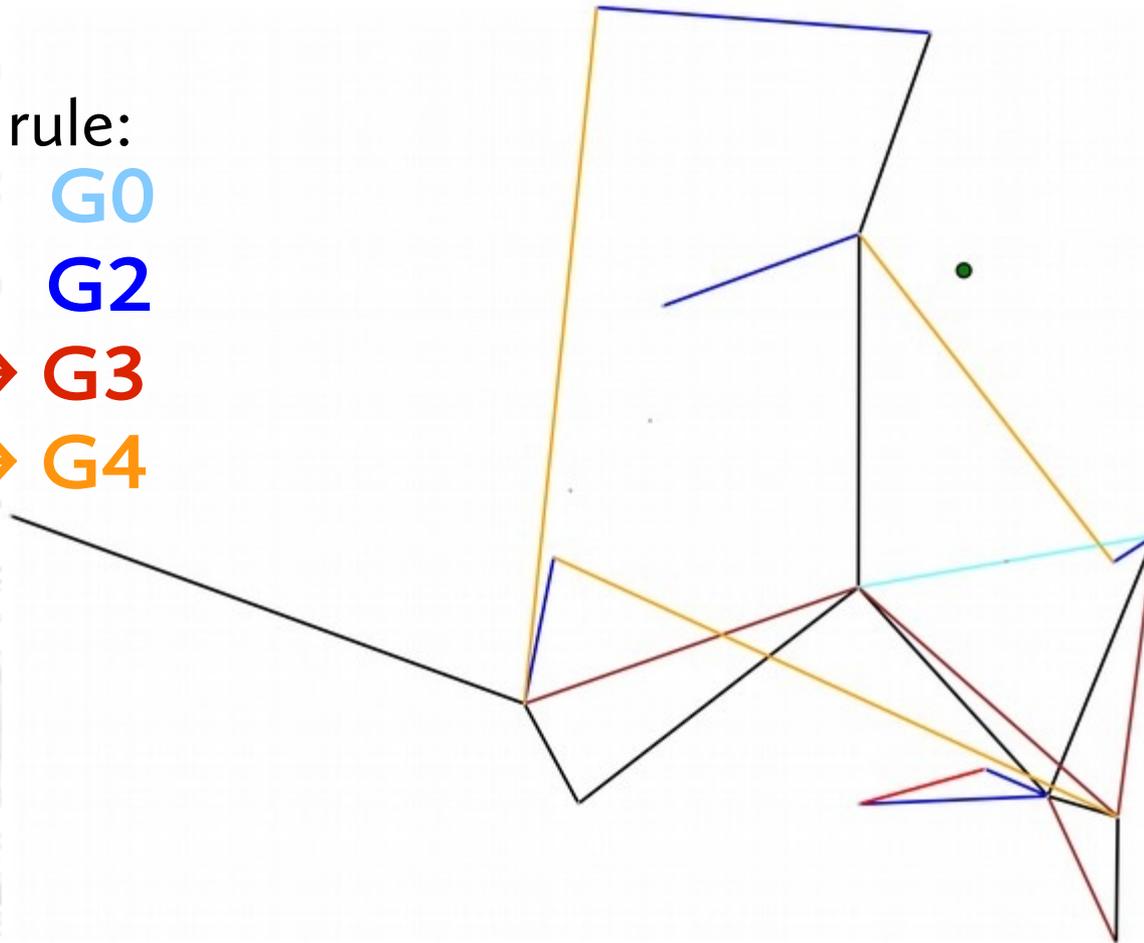
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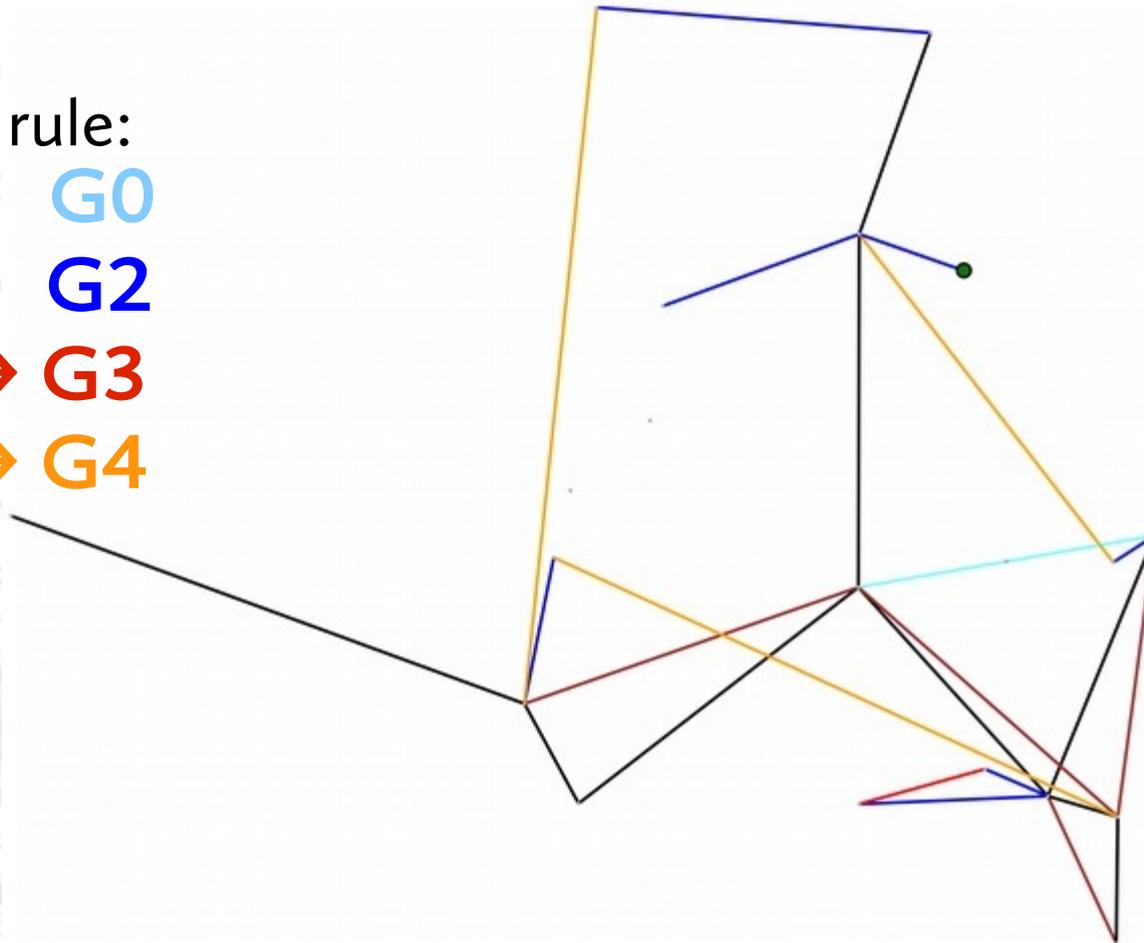
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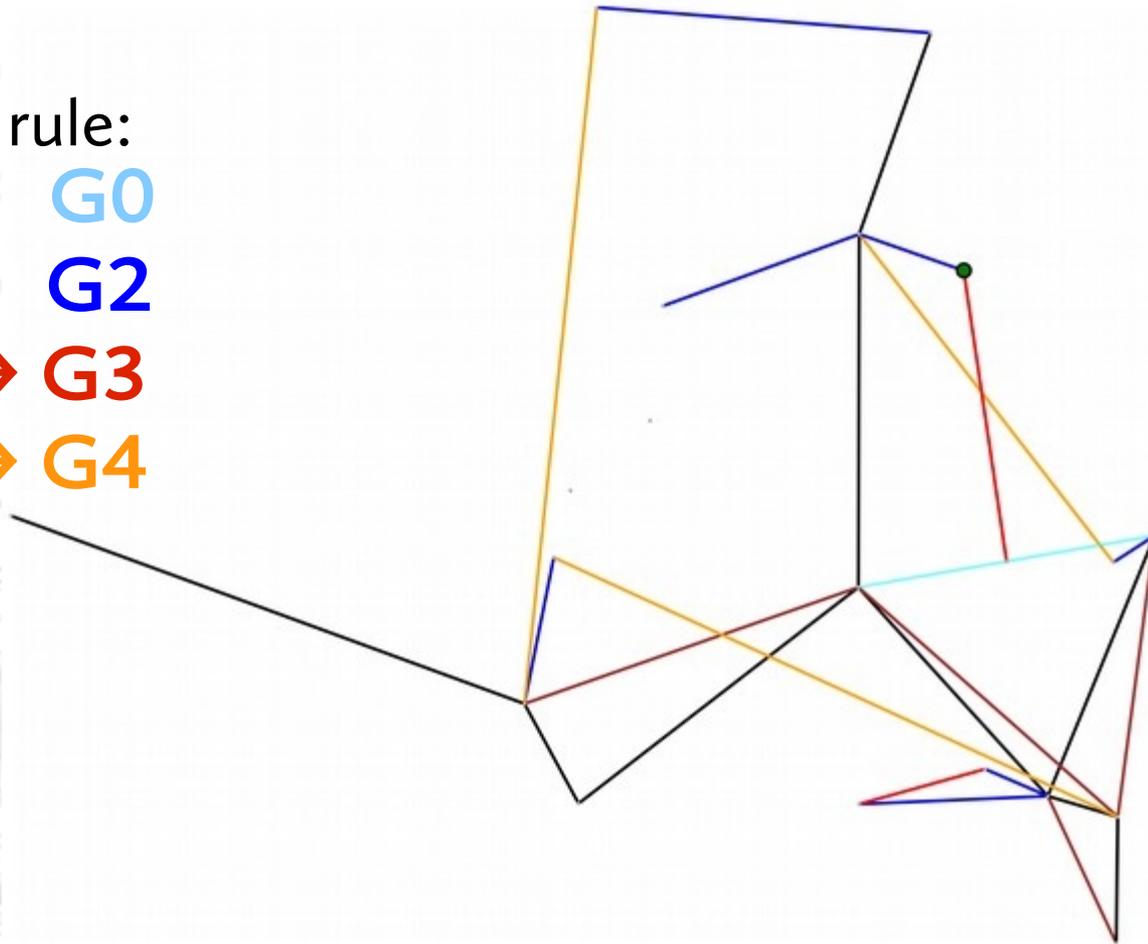
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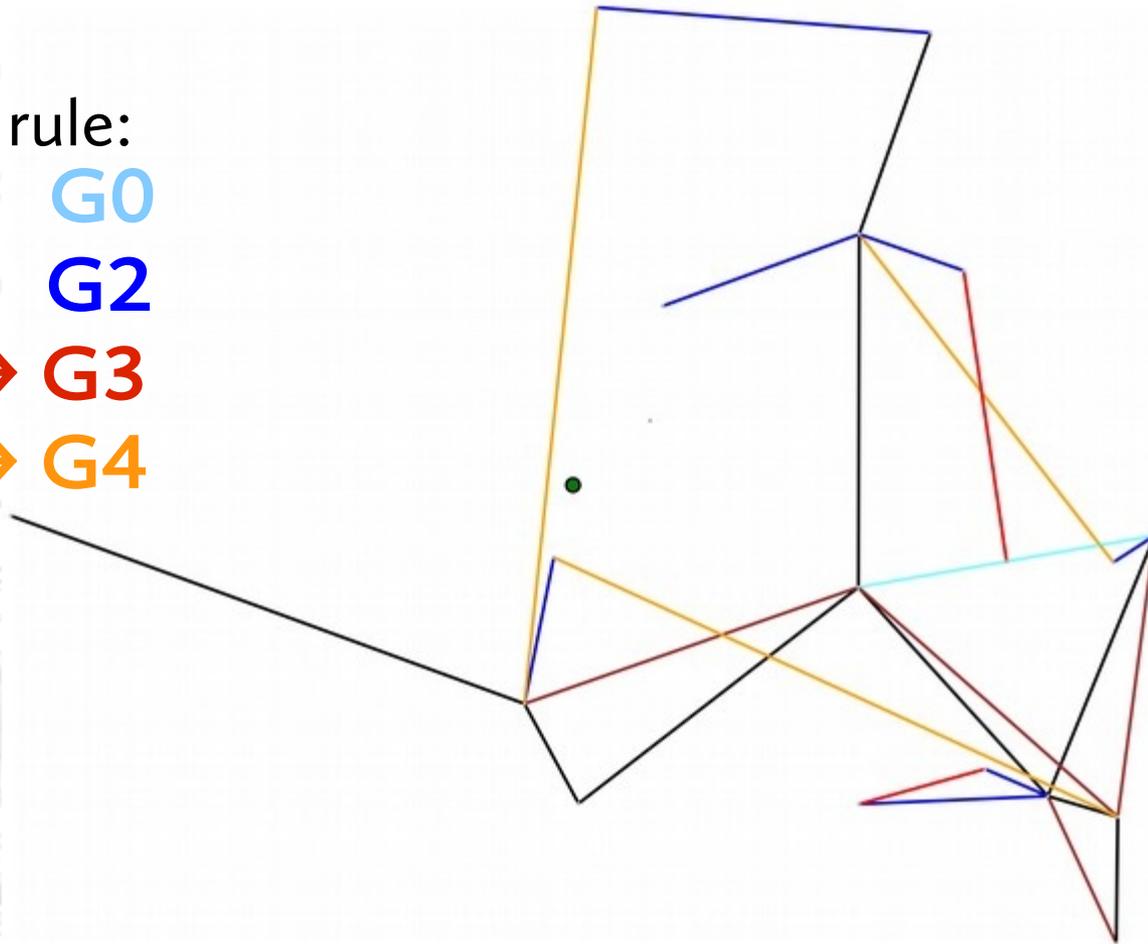
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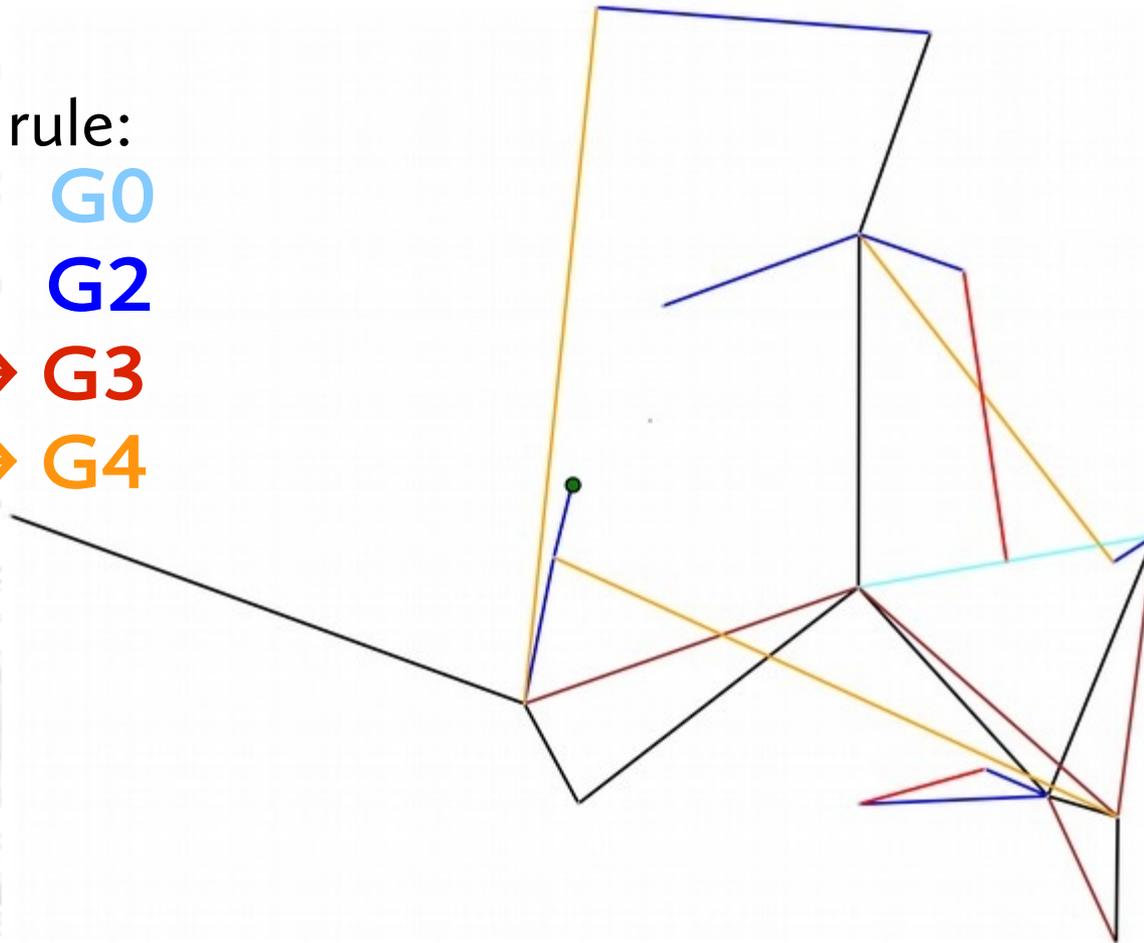
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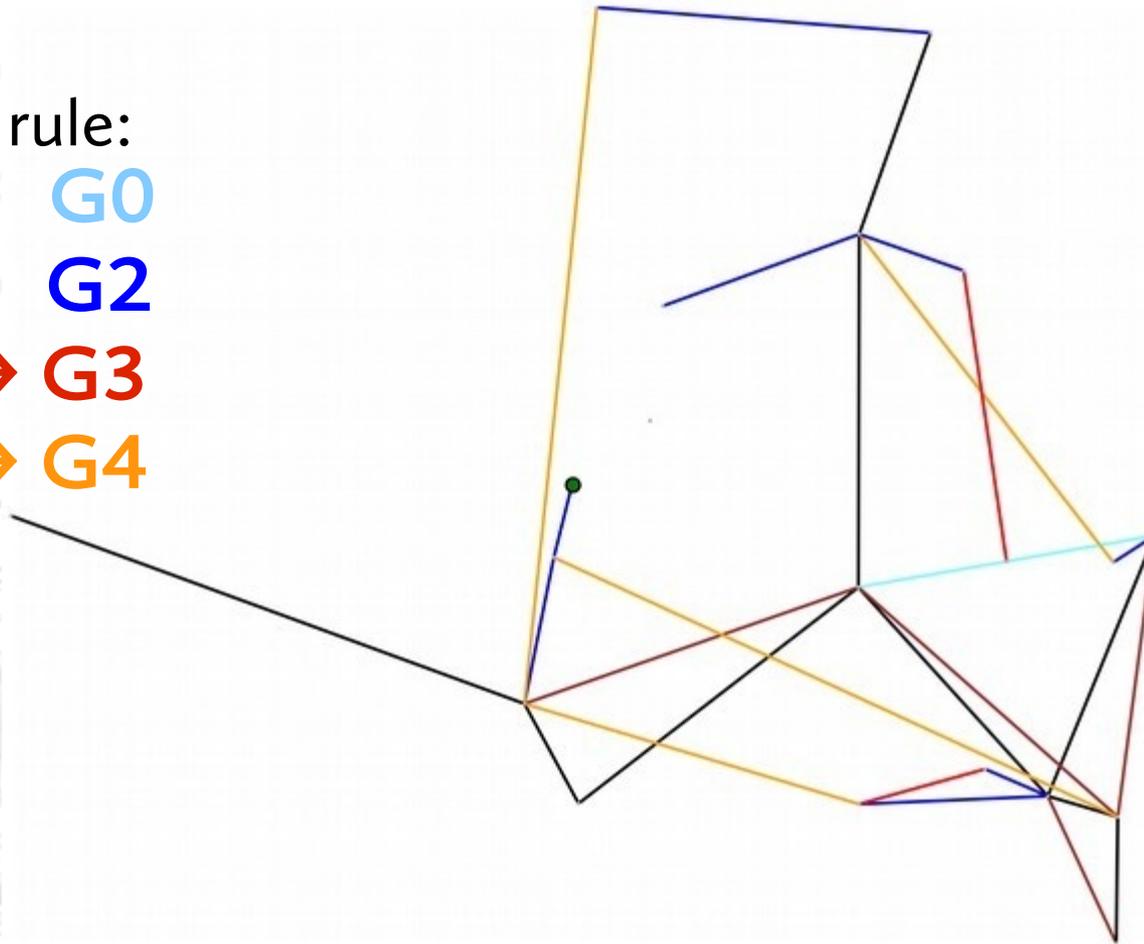
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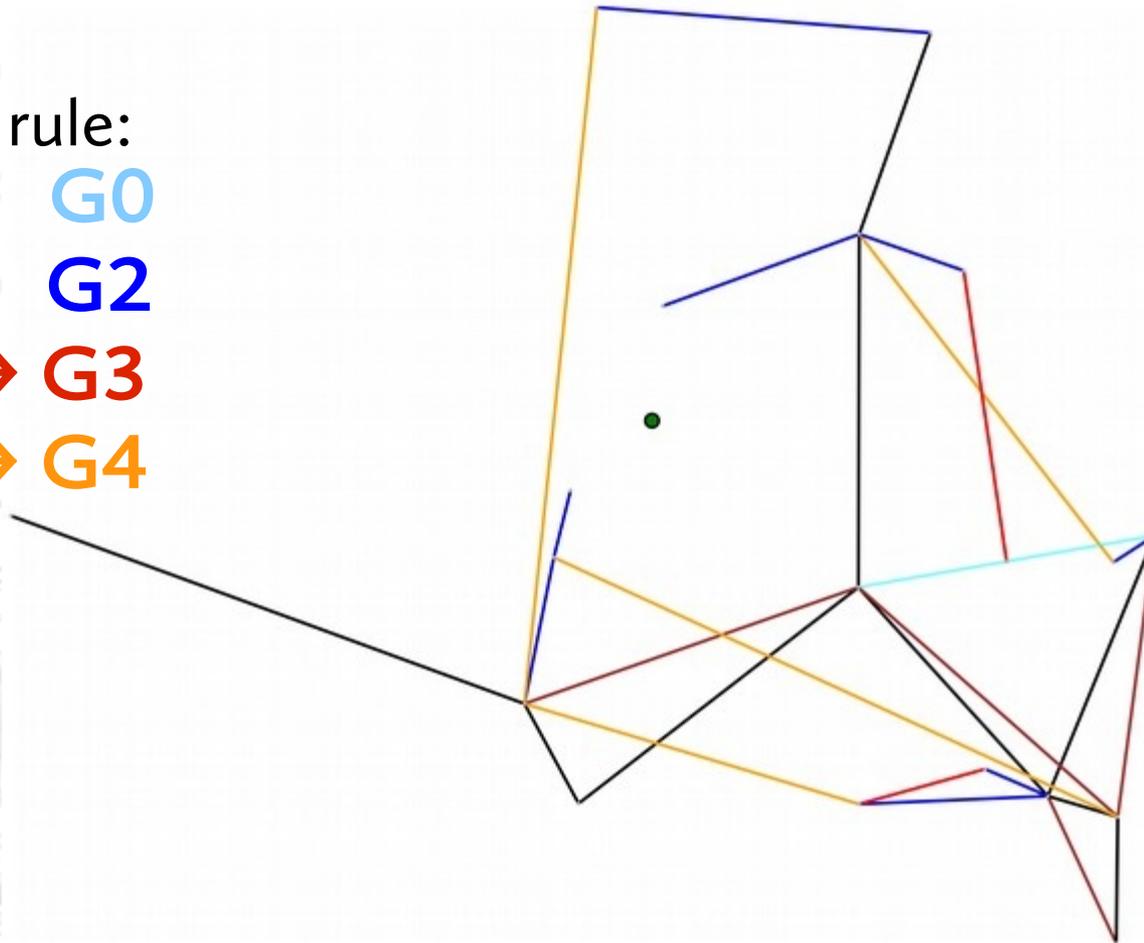
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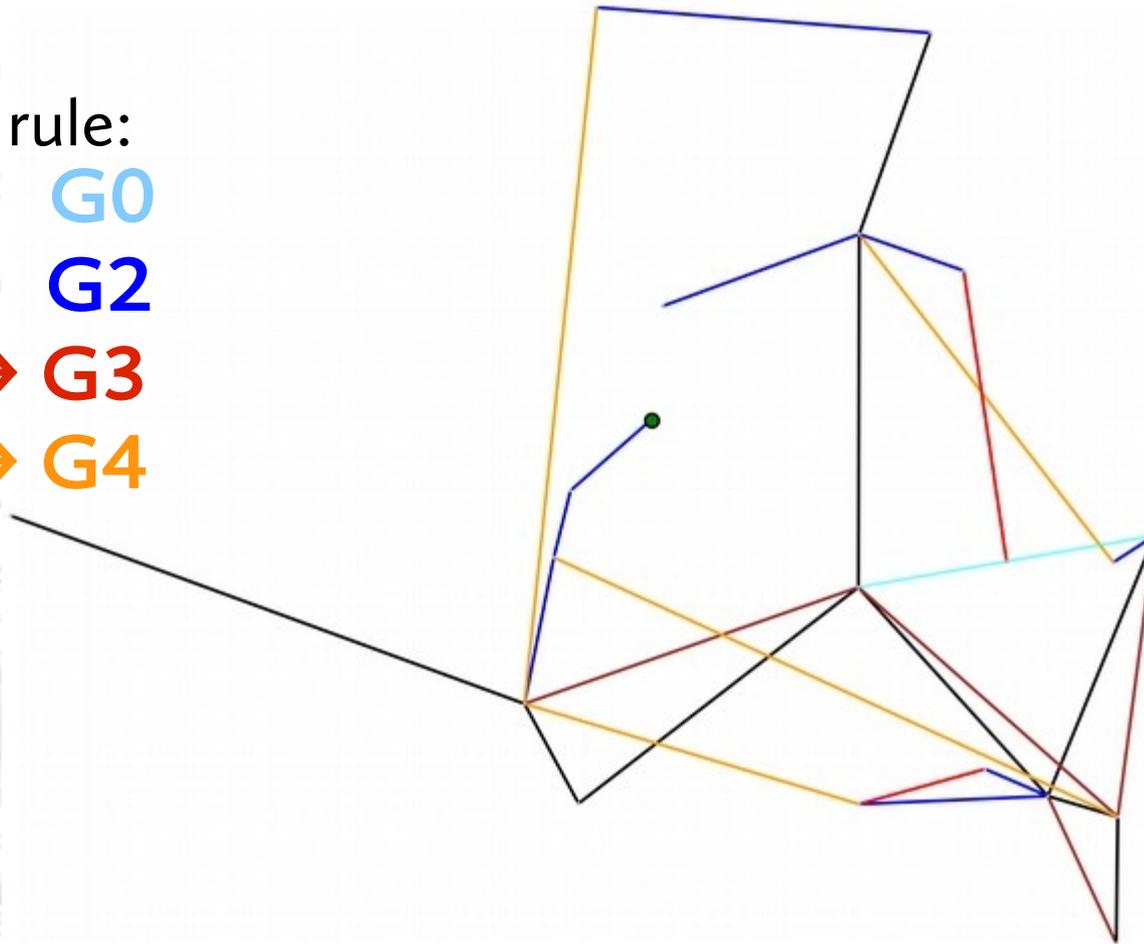
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Ref.: Schultz et al., EPJ ST, 2014

3. Ensemble Analysis



Power Grid Ensemble

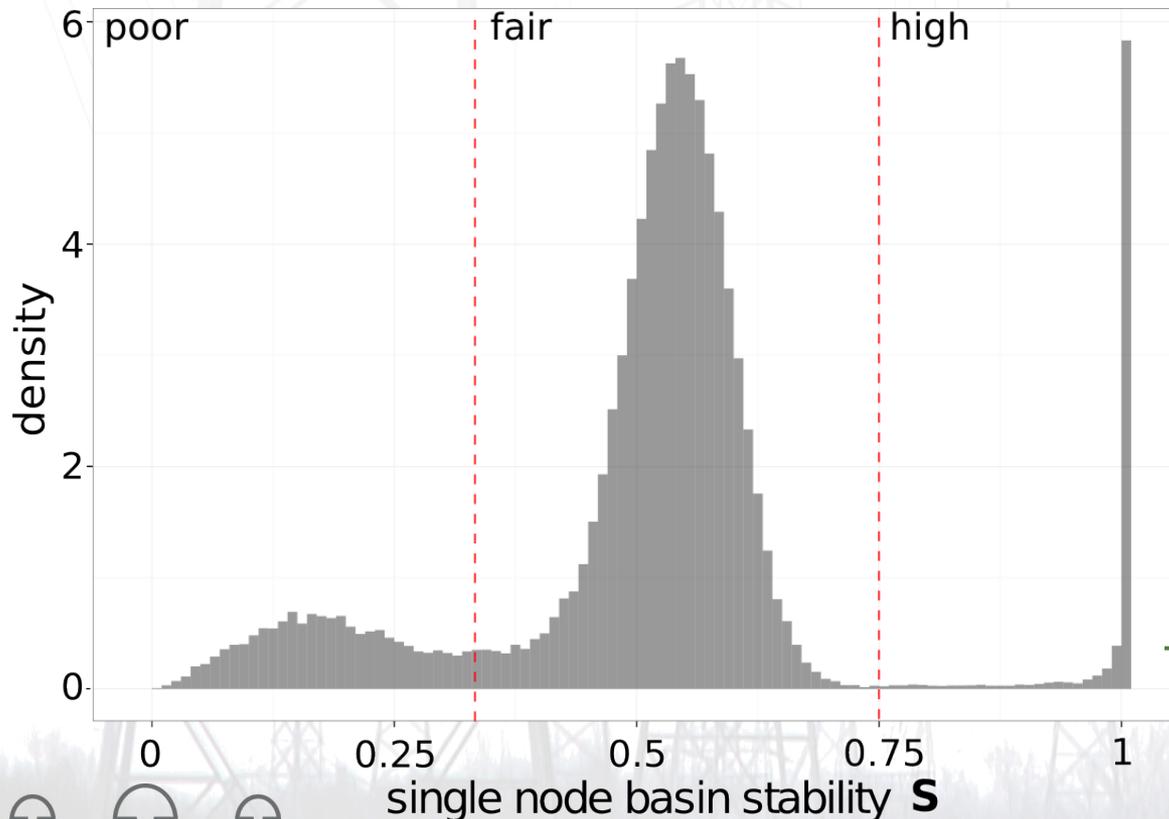
~600 networks
100 nodes each
 $\bar{k} = 2.7$

$$\ddot{\phi}_i = P_i - \alpha_i \dot{\phi}_i - \sum_{j=1}^N K_{ij} \sin(\phi_i - \phi_j)$$



realistic choice:

$$K_{ij} \propto 1/d_{ij}$$



→ classification of SNBS

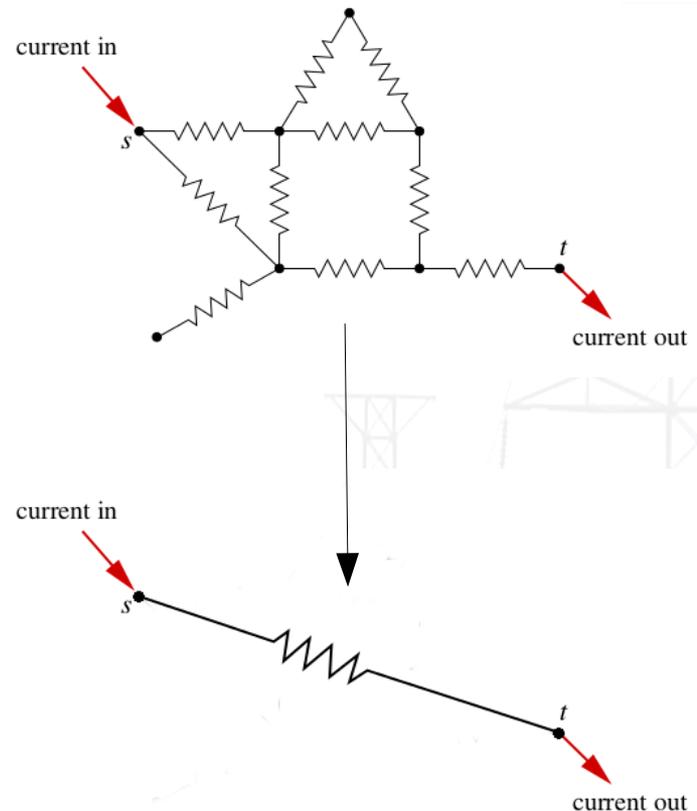
Effective Resistances

vs. network distance

- resistance of a **single edge**, virtually replacing all paths **between s and t** in an equivalent circuit
- **replace path-based observables**

$$\text{Exp. } ERCC_i = 1 / \langle ER_{ij} \rangle_j$$

$$ER_{ij} = (b_i - b_j)^\top \mathbf{L}^{-1} (b_i - b_j) = L_{ii}^{-1} - L_{ij}^{-1} - L_{ji}^{-1} + L_{jj}^{-1}$$

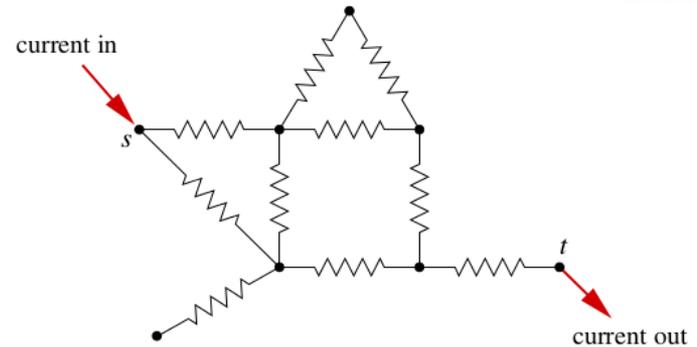


Ref.: Klein&Randic 1993, Dörfler&Bullo 2010

Newman's Current Flow Betweenness

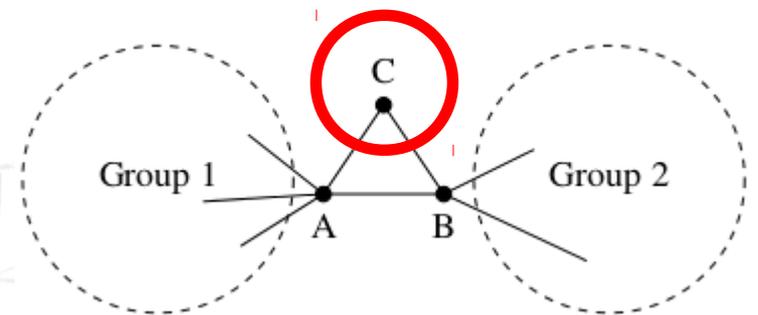
vs. shortest path betweenness

- Based on virtual current through node i according to Kirchhoff's laws
- Generalised for arbitrary admittance



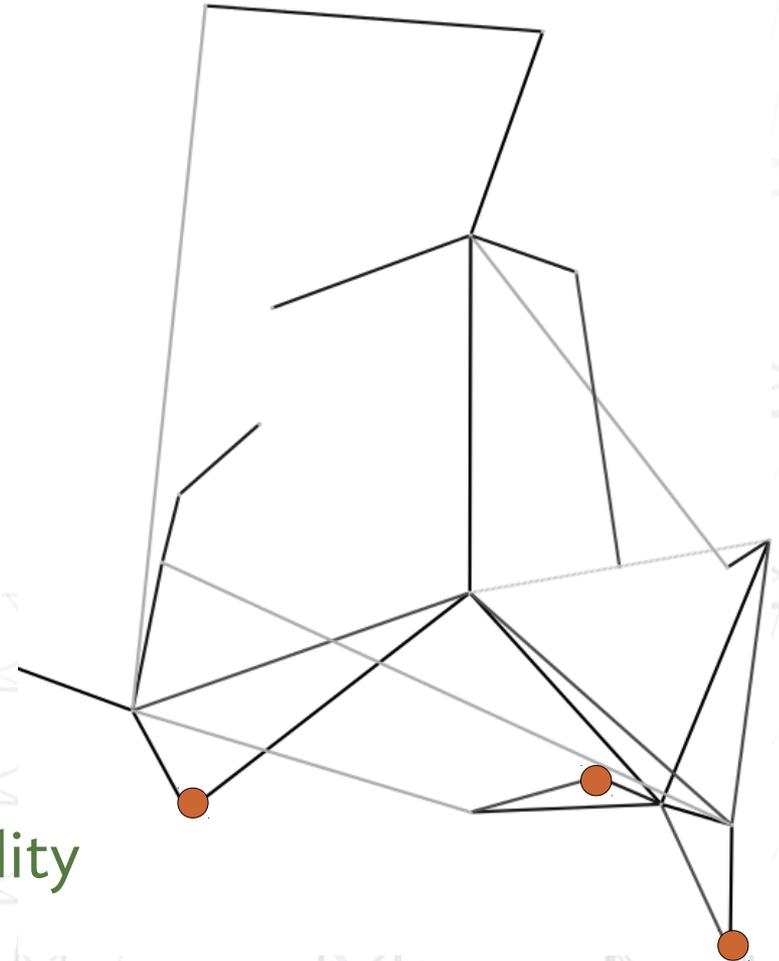
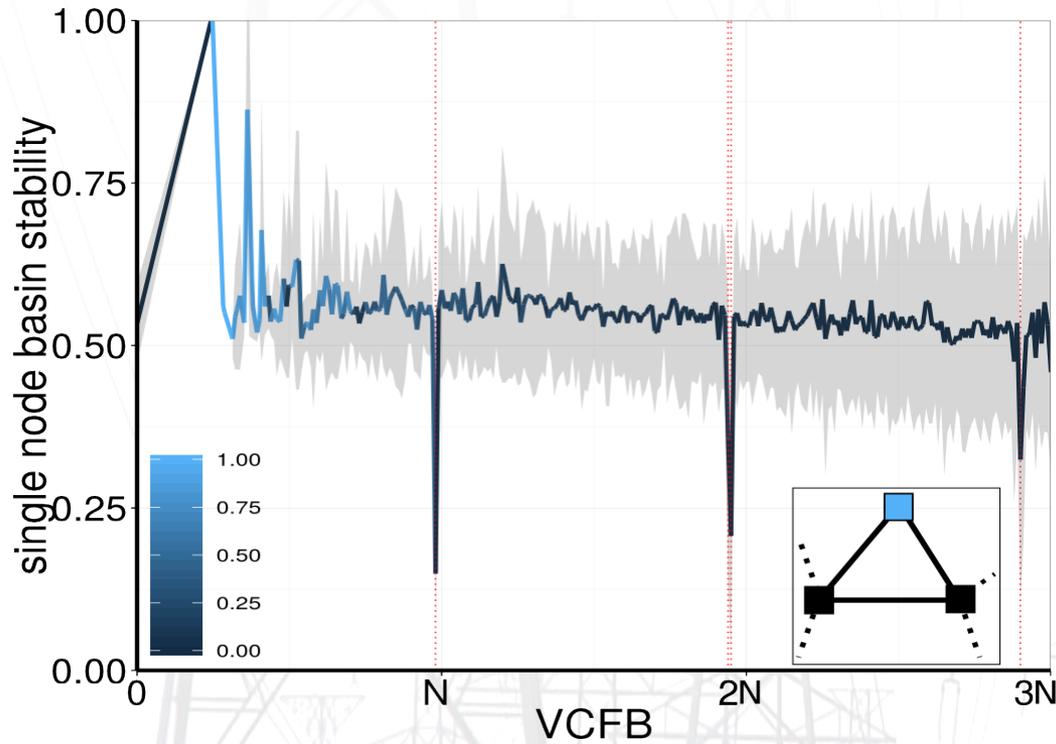
$$VCFB_i = \frac{2}{n(n-1)} \sum_{s < t} I_i^{st}$$

$$I_i^{st} = \frac{1}{2} \sum_j Y_{ij} |V_i - V_j|$$



Ref.: Newman, Social Networks (27), 2005

Nodes on Detours

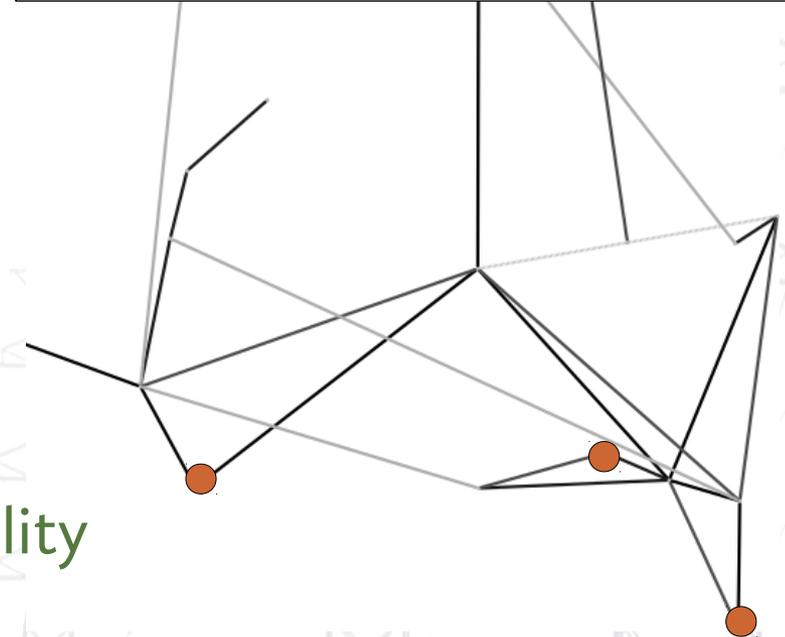
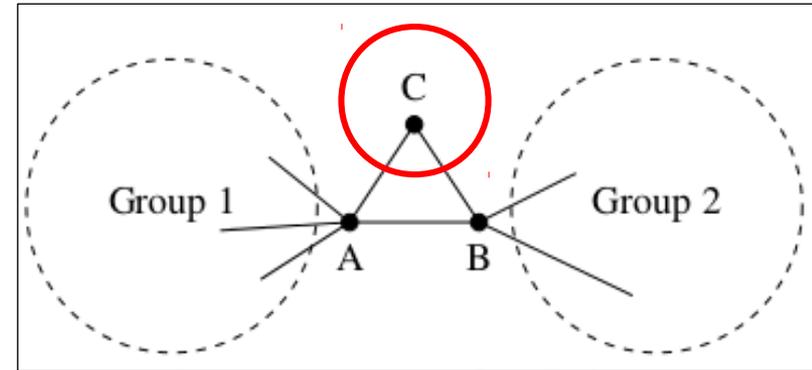
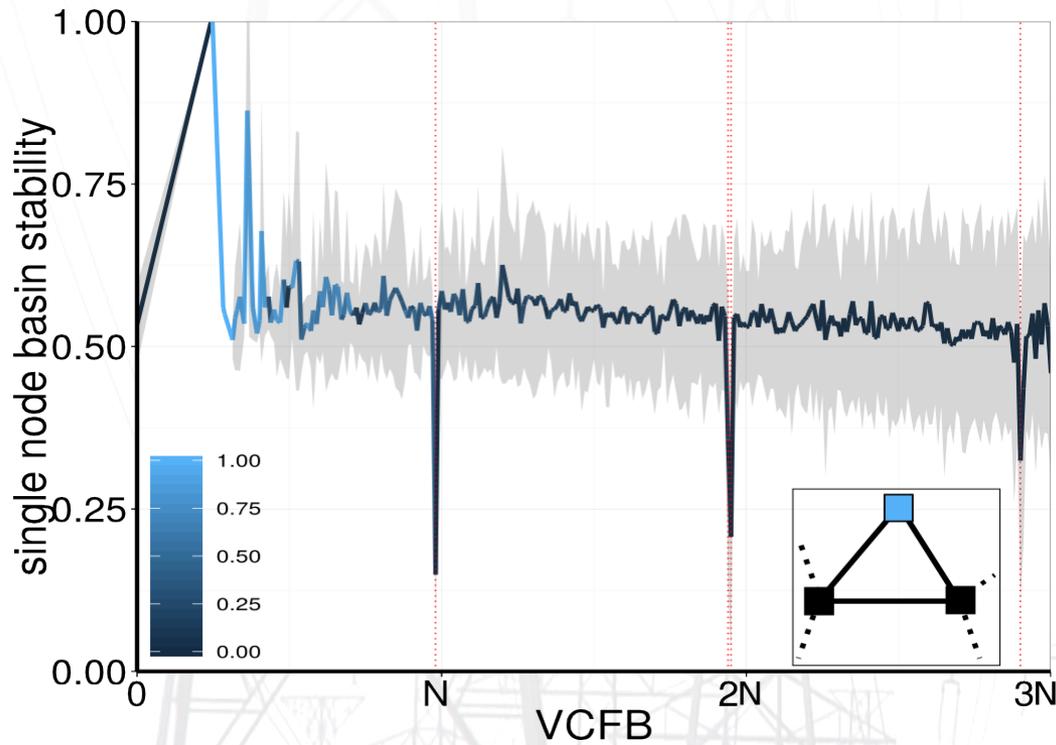


- Indicates fair to high basin stability
- Identified using VCFB

Ref.: Newman, Social Networks (27), 2005

Ref.: Schultz et. al., NJP 16, 125001 (2014)

Nodes on Detours

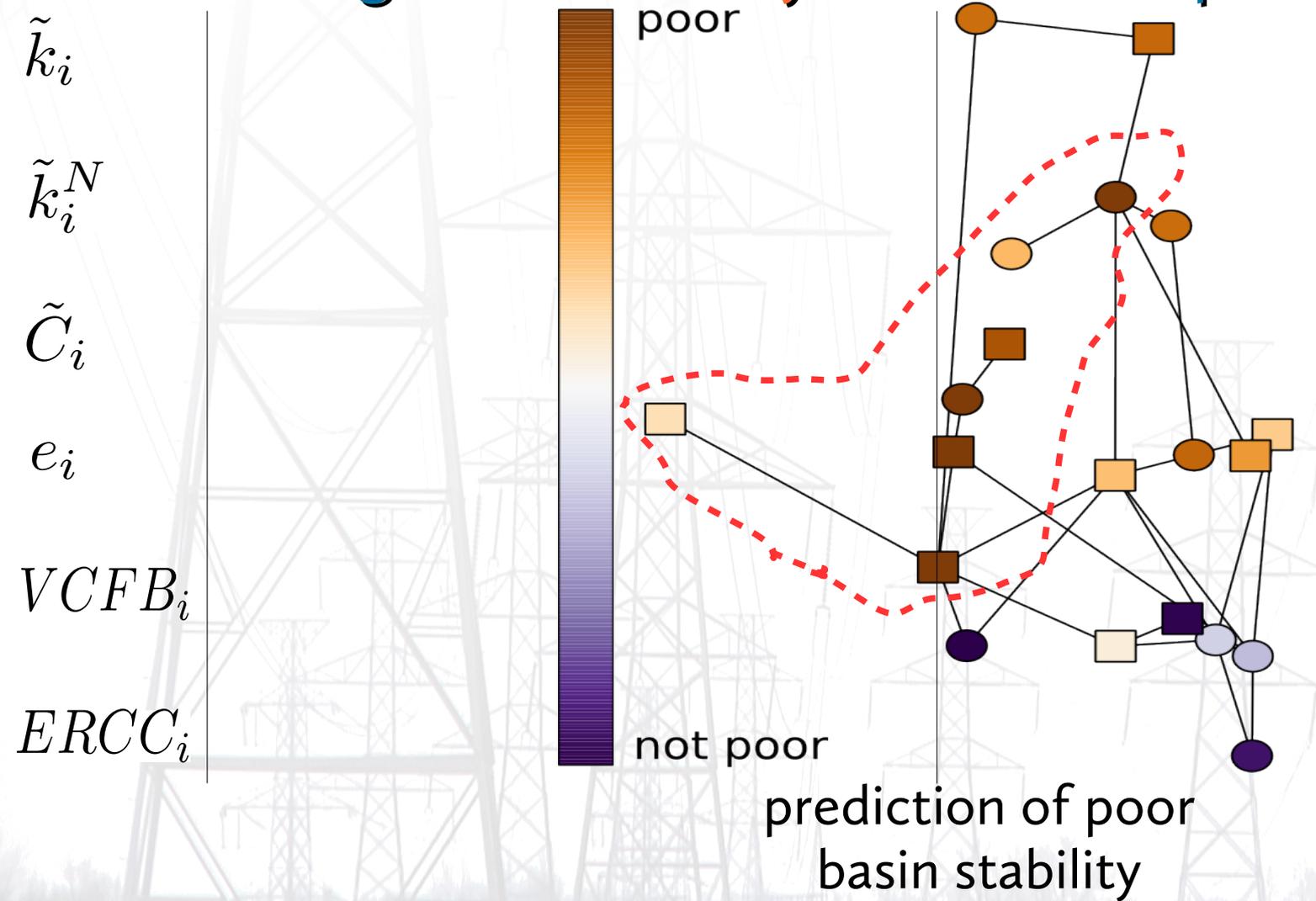


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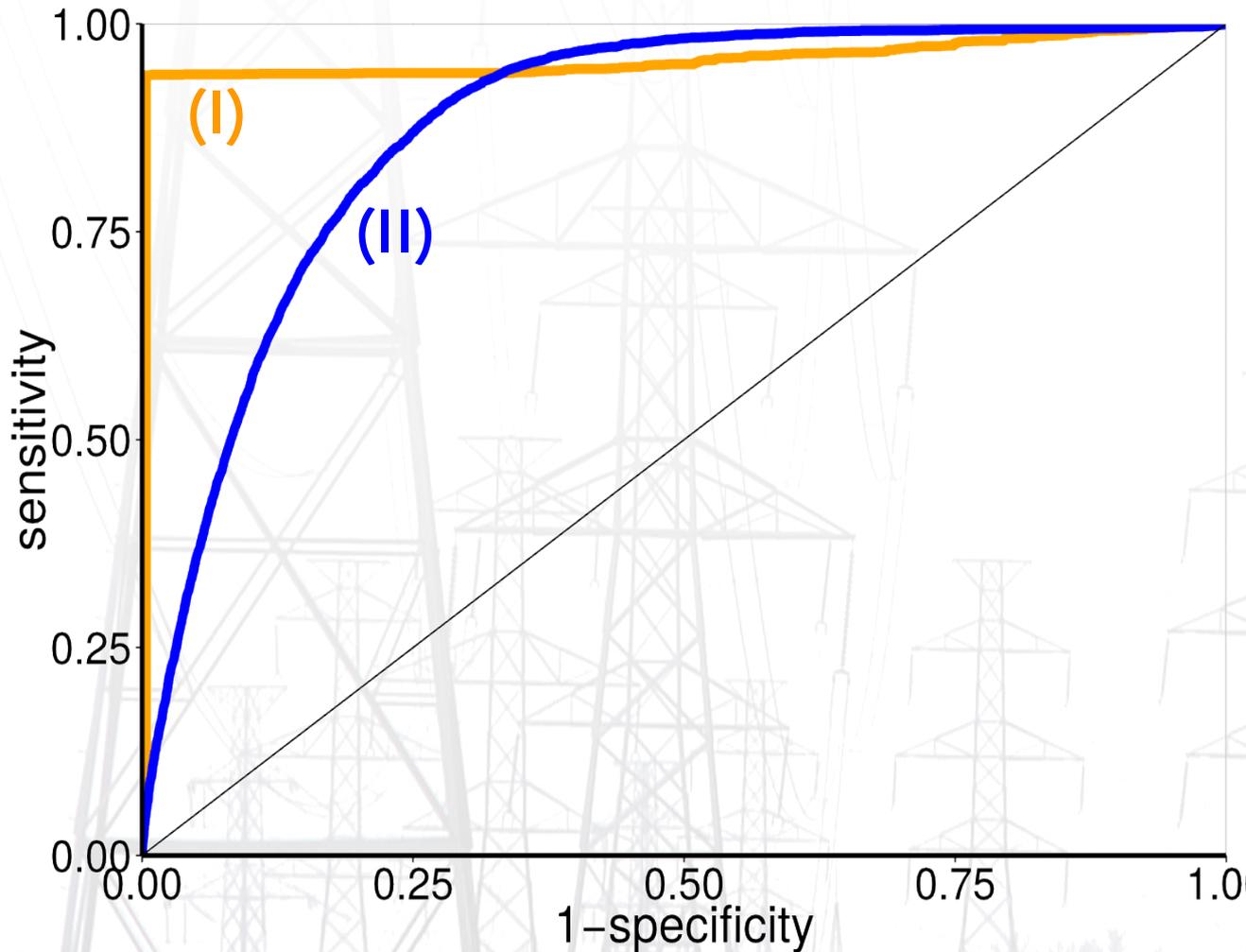
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Predicting Poor Stability from the Topology



Receiver Operating Characteristic



- (0) detour nodes
- (I) dead ends
Sen: 94%
Spec: 99%
- (II) all others
Sen: 84%
Spec: 77%

Ref.: Schultz et. al., NJP 16, 125001 (2014)

Take Home Messages

Random topology model capable to create realistic power grid networks

Spatial information to estimate model parameters, surrogate networks for hypothesis testing, ensemble analysis etc.

Effective resistances and current-flow betweenness reveal additional information for electrical networks

It is possible to predict the weak points using only structural information

Dead ends diminish while detour motifs enhance network resilience

Publications

Detours Around Basin Stability in Power Networks

Paul Schultz, Jobst Heitzig, and Jürgen Kurths
New Journal of Physics 16 (2014) 125001
DOI: 10.1088/1367-2630/16/12/125001



A Random Growth Model for Power Grids and Other Spatially Embedded Infrastructure Networks

Paul Schultz, Jobst Heitzig, and Jürgen Kurths
Eur. Phys. J. Special Topics on "Resilient power grids and extreme events" (2014)
DOI: 10.1140/epjst/e2014-02279-6

