

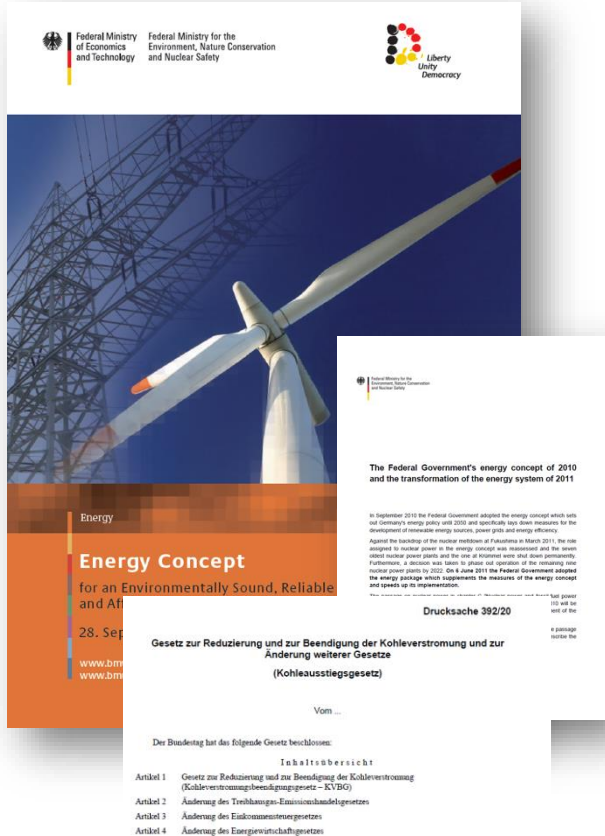


QUANTUM ANNEALING BASED POWER GRID PARTITIONING FOR PARALLEL SIMULATION

PROF. DR.-ING ANDREA BENIGNI

TRANSFORMATION OF THE ENERGY SYSTEM

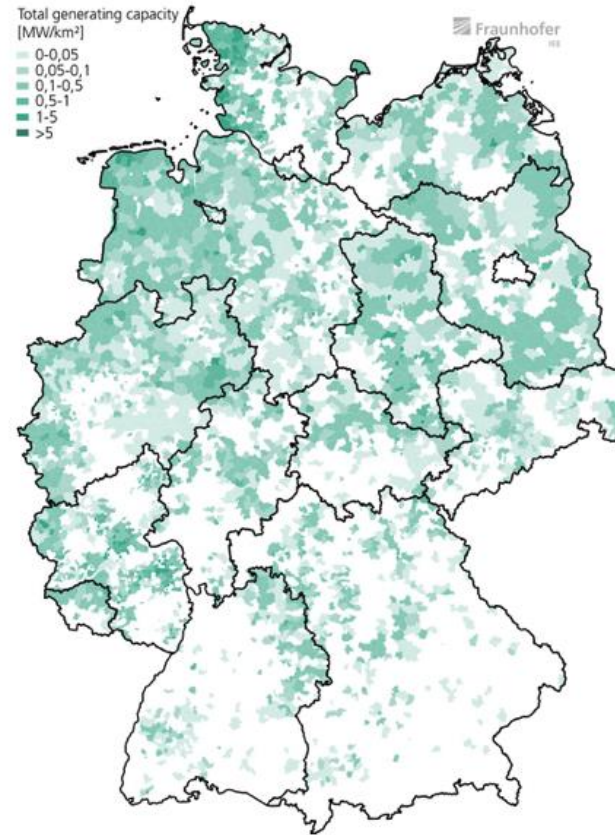
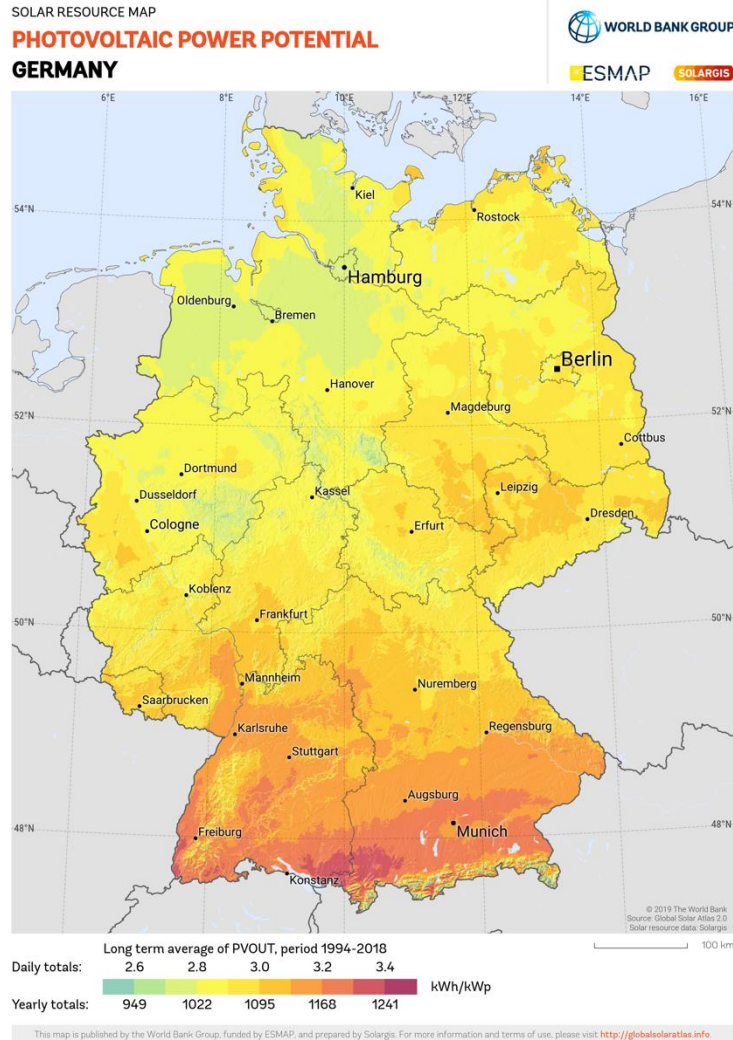
The Energy Transition Dilemma - The German Energiewende in a Changing World



- Reduction of greenhouse gas emissions by 40% by 2020, 55% by 2030, and at least 80% by 2050 compared to 1990 levels?!
 - Phase out of nuclear power by the end of 2022 (stretch operation until April 2023 in planning)?!
 - Phase out of hard coal and lignite by the end of 2038?!
 - Abandonment of gas from Russia as a bridging energy source!
- How can we still ensure a reliable, affordable and sustainable energy supply?
- How can we handle the complexity of designing and operating resilient, more decentralized & sector-integrated energy systems?

THE TRANSFORMATION OF THE ENERGY SYSTEM

“Energiewende” started as a “Stromwende”

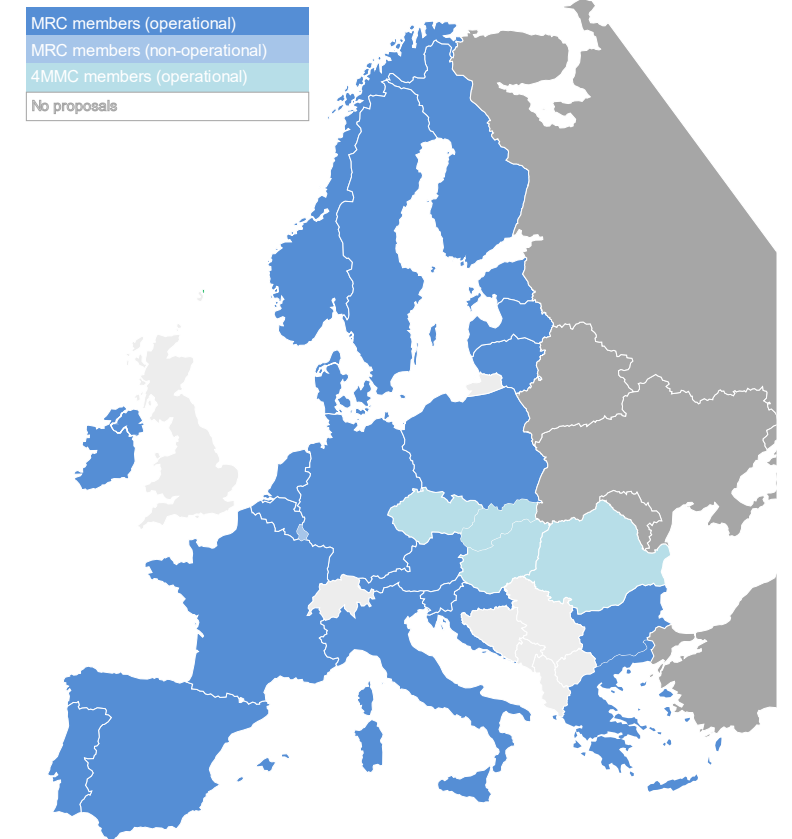


Exact postcode distribution of installed onshore wind capacity in Germany in 2018

COMPLEXITY OF LARGE-SCALE SYSTEMS

The number of actors in the market is increasing strongly

- The European electricity market is a multiscale system that couples 28 countries representing 95% of the EU electricity demand
 - The market coupling determines dispatch of generation, loads, prices and cross-border power flows
 - Security of supply is secured by the n-1 criteria
- For the analysis of modern energy systems, both the continental and the local level must be considered in a multiscale fashion

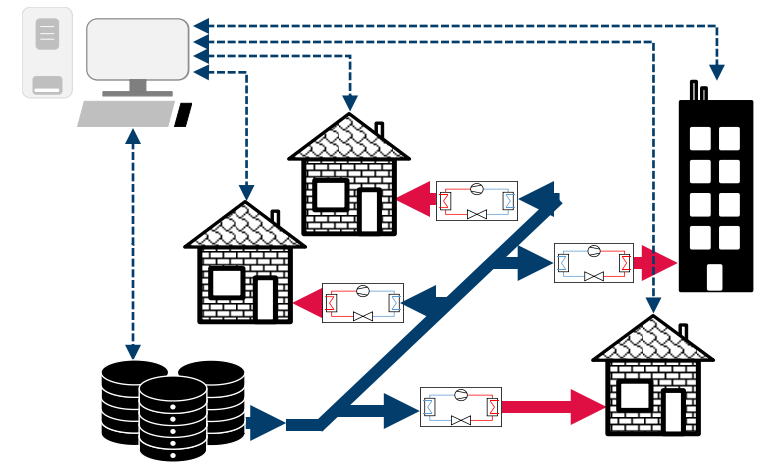
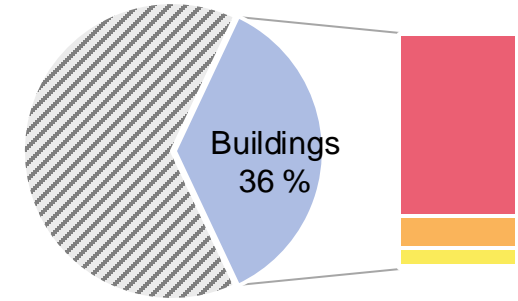


Source: https://www.entsoe.eu/network_codes/cacm/implementation/sdac/#sdac-geographical-scope-and-extensions

THE TRANSFORMATION OF THE ENERGY SYSTEM

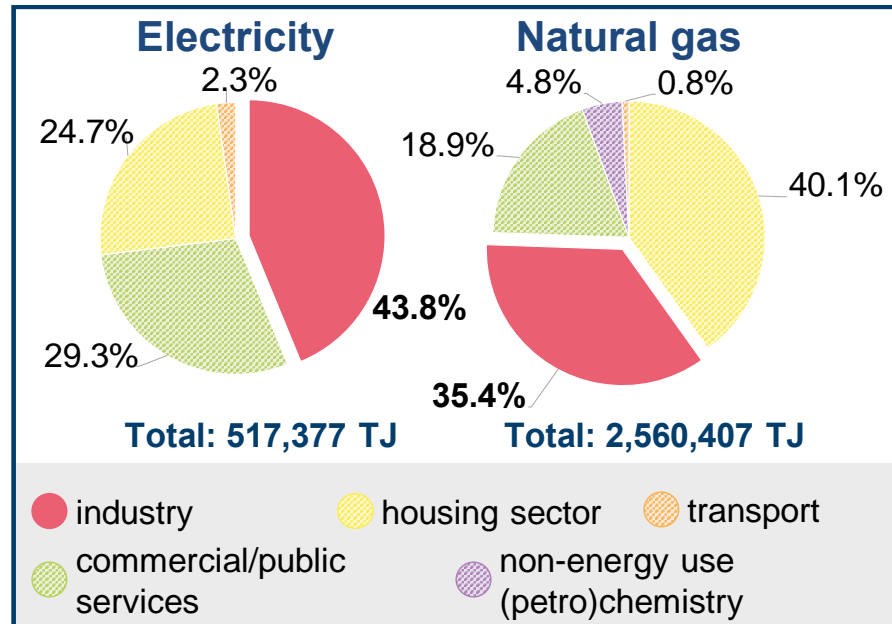
“Energiewende“ and building sector

- Lowering energy demand and raised user awareness in buildings
(while fulfilling higher comfort demand / cooling demand)
- Buildings can offer flexibility options to the electricity grid, e.g. by increased share of heat pumps in new buildings
- Modern district heating networks gain importance
 - Existing building stock aged (62 % built before 1979)
 - Gradual phase out of fossil plants
 - Integration of waste heat and renewables heat sources



INDUSTRIAL SYSTEMS

German Energy Demand* in 2016^[1]



*relative to final energy demand

[1] International Energy Agency. *World Energy Statistics 2018*. Paris: IEA Publications, 2018.

Industry is responsible for a **large share of energy demand** (>30% of final energy for natural gas/electricity).

Decarbonization: Industry needs to satisfy its energy demand from **CO₂-neutral sources**.

To achieve cost competitiveness, **sector integration** and **demand side management** will be key.

THE DECARBONISATION OF THE STEEL INDUSTRY

Handelsblatt

STAHLHERSTELLUNG

CO2-freie Produktion bis 2050: Thyssen-Krupp will Hochöfen dichtmachen

von: Kevin Knitterscheidt
Datum: 21.01.2019 04:15 Uhr

Thyssen-Krupp will bis 2050 zehn Milliarden Euro in die CO2-freie Stahlerzeugung investieren. Für die Branche beginnt ein Rennen gegen die Zeit.



CO2-Verursacher

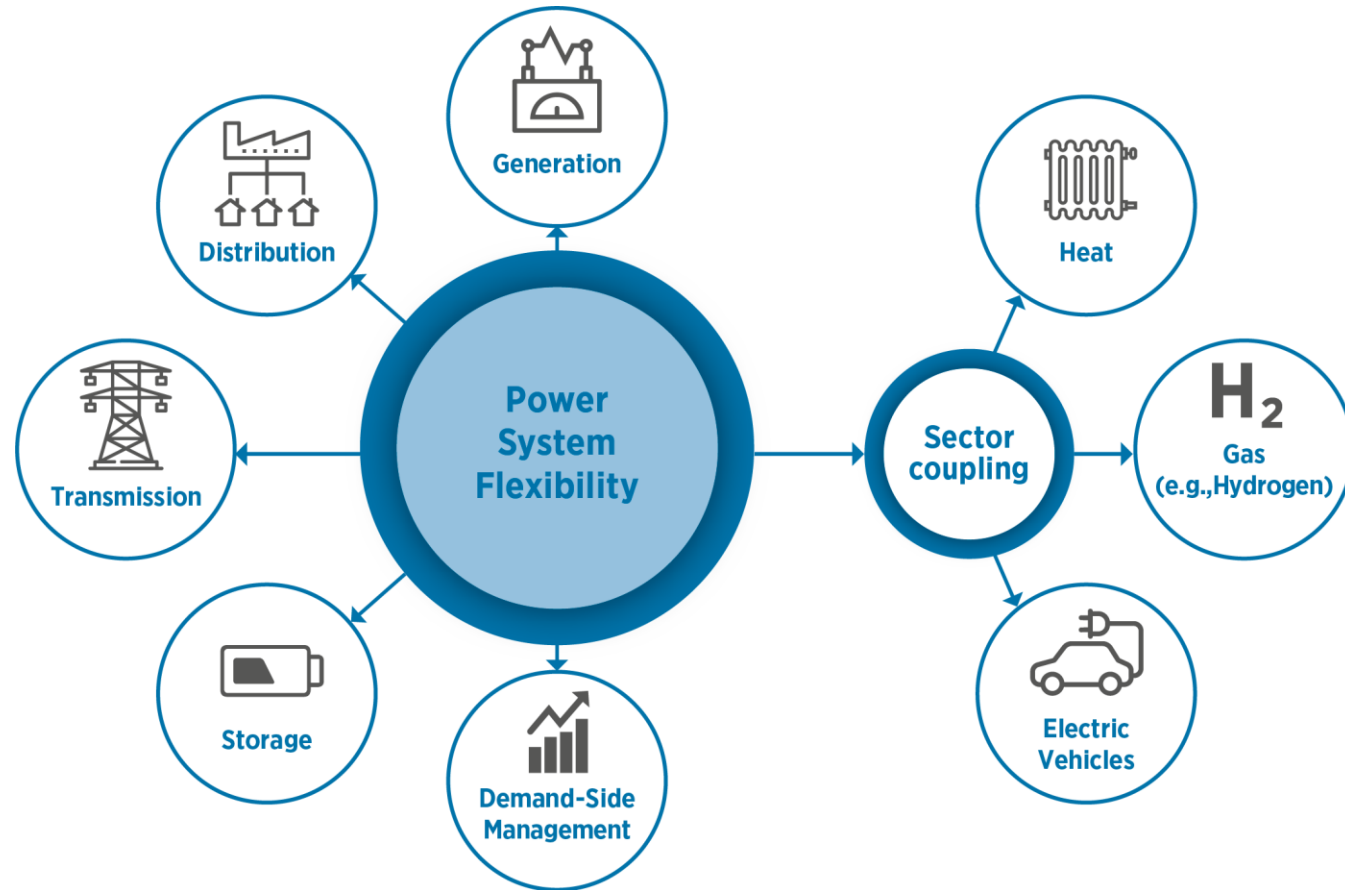
Die Stahlbranche denkt langsam um.

(Foto: dapd)

- It is the industry's task to make its contribution to climate protection, said Thyssen Krupp Steel Director Andreas Goss
- The company plans to invest a total of ten billion euros over the next 30 years in the modernization of its processes in order to replace coking coal.
- Over the next few decades, plants will be gradually modernized and rebuilt until the 2050 Duisburg blast furnaces are completely replaced by direct reduction systems and electric furnaces.

FLEXIBILITY VS. SECTOR COUPLING

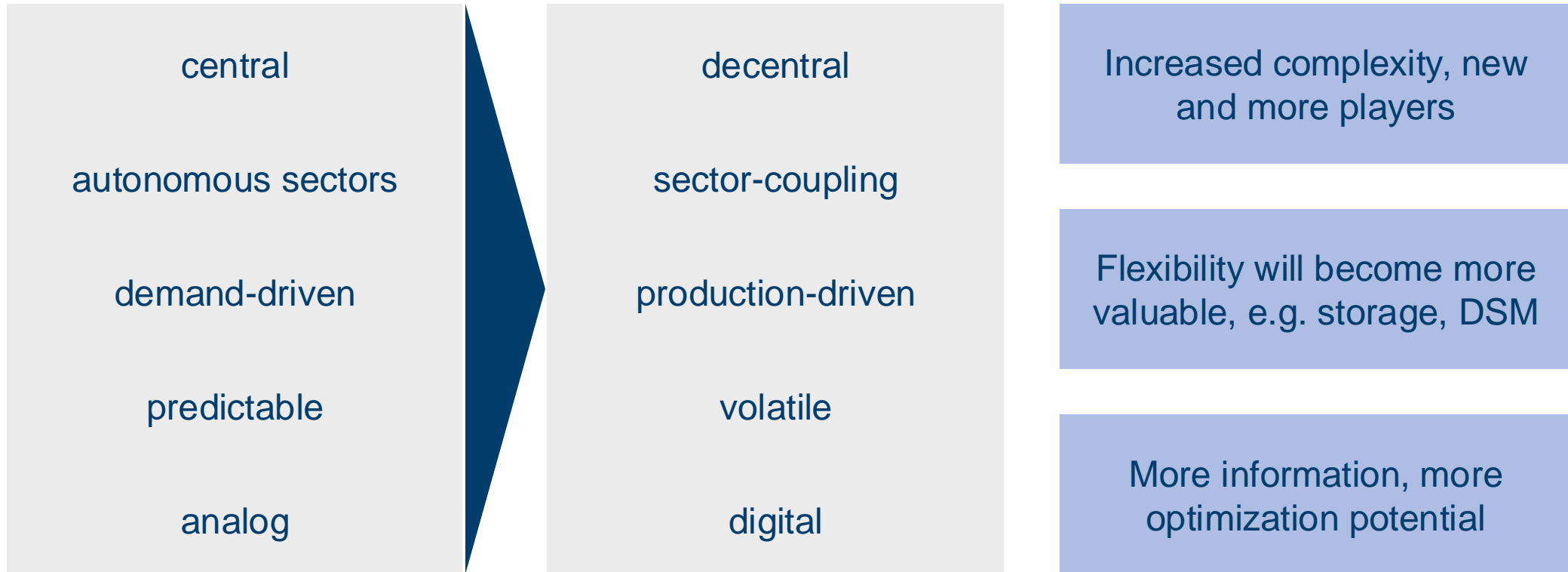
Autonomous → sector coupling



Flexibility is the capability of power system to maintain its operation under uncertainties.

Sector coupling can offer flexibility to the power system with the focus is on the decarbonization of other sectors.

TRANSFORMATION OF THE ENERGY SYSTEM

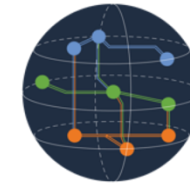


COMPUTATIONAL CHALLENGES

- Need for longer simulation horizon while increasing temporal resolution (due to the reduced system inertia while introducing PtX technologies and seasonal storage)
- Need for spatial disaggregation (due to higher interconnections and distributed generation penetration)
- Need for higher number of contingencies and scenarios to considered (due to uncertainties and control dependencies)

Can quantum computing supports power system?
For which applications?

QUANTUM-BASED ENERGY GRIDS



QuGrids

Rethinking Planning and Operation of Energy Grids

Planning and Design

Simulation and optimization are used to support topology and technology selection, components sizing, operating modes definition.

Existing approaches limited by computational complexity (Spatial and temporal span, Sector coupling)



Quantum Computing for Energy Grids

Which computational challenge would benefit from QC?
How to combine quantum and traditional computing ?
Which impact would this have on systems metrics?

Operation and Automation

Digital platform and ICT infrastructure used to monitor system operation, schedule assets (e.g. generators and storage), enable flexibility services

Existing approaches limited by the amount of involved players and generated data, latency and bandwidth, security and privacy concerns



Quantum Communication for Energy Grids

How can QKD schemes be used in energy grid ?
Can we design a fully quantum-based automation architecture ?

QUANTUM-BASED ENERGY GRIDS



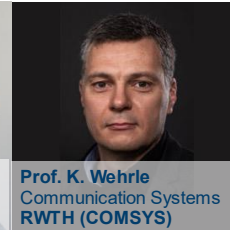
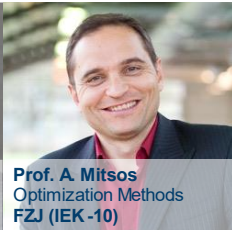
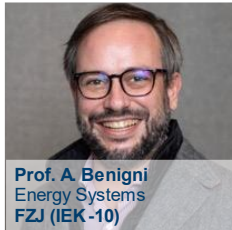
Rethinking Planning and Operation of Energy Grids

Develop first groundbreaking results so to highlight the value of quantum technologies, identify the shortcomings and define a research and development roadmap.



Develop the conditions for a long-term operation of the QuGrids consortium

Create new teaching and training material for workforce development

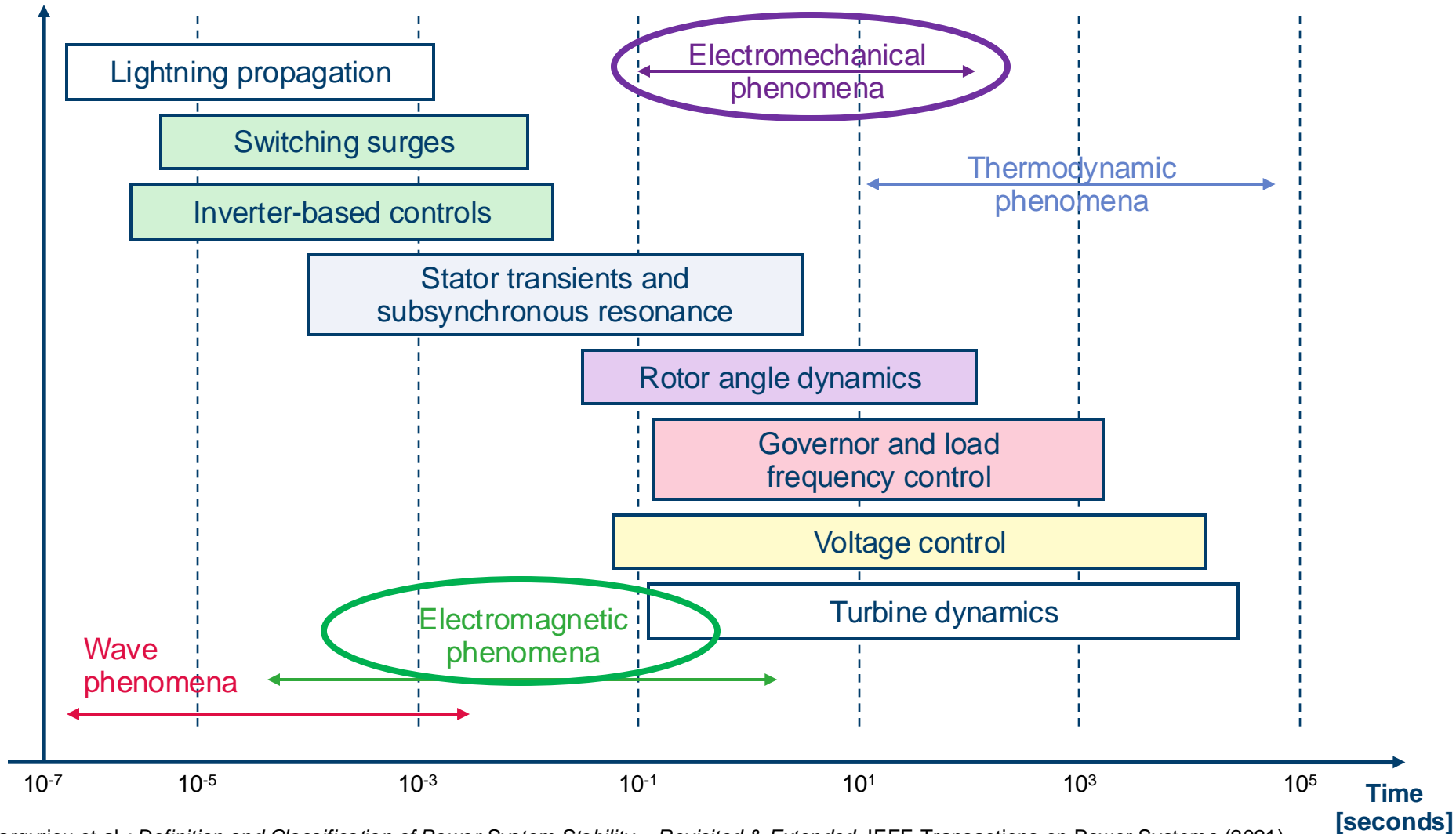


DEMO CASES FOR QUANTUM-BASED ENERGY GRIDS

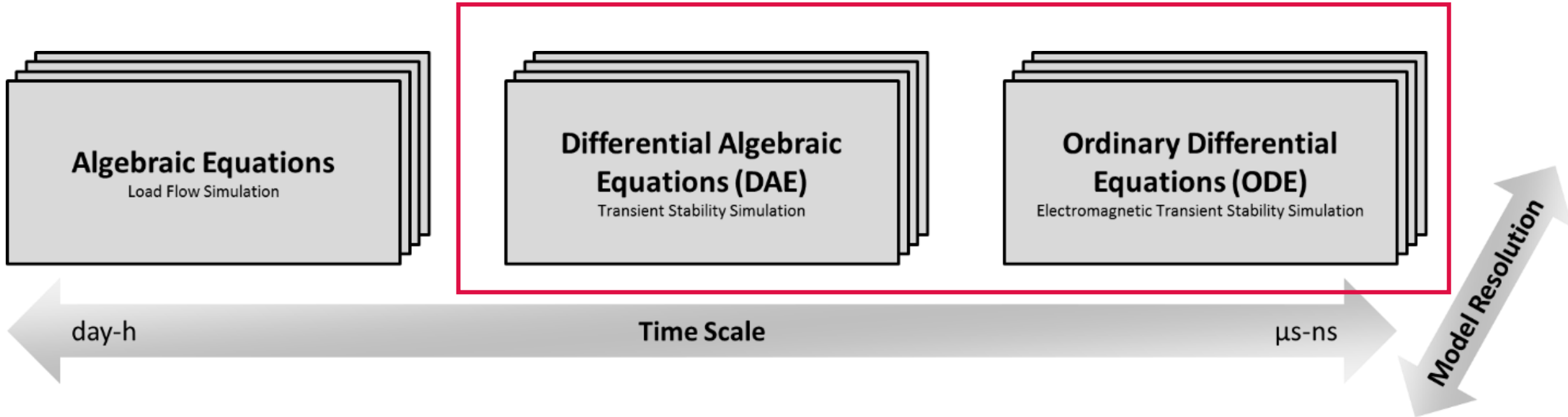
Objective: Develop first groundbreaking results so to highlight the value of quantum technologies for energy grids and to identify the shortcomings of existing quantum technologies so to define a research and development roadmap.

- WP1.1 Power System Monitoring
- WP1.2 Optimal Planning and Operation of Multi-Carrier Energy Networks
- WP1.3 Static and Dynamic Simulation Under Uncertainty
- WP1.4 Quantum based edge grids device
- WP1.5 Quantum communication in energy grids

POWER SYSTEM TIME SCALES

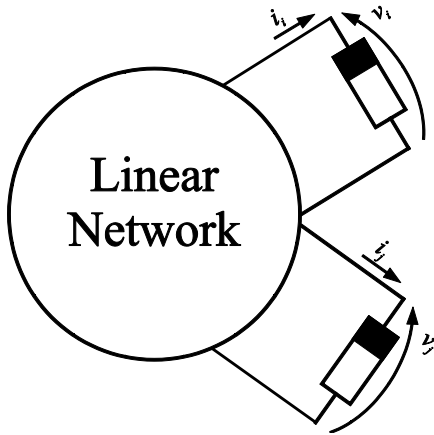


POWER SYSTEM SIMULATION



POWER SYSTEM SIMULATION

A parallel approach



$$\frac{di_i^n}{dt} = f(v, i, x_i^n, u_i^n, t)$$

$$I_i^n(k+1) = f(v(k), i(k), x_i^n(k), u_i^n(k), k)$$

$$\frac{dv_j^n}{dt} = f(v, i, x_j^n, u_j^n, t)$$

$$V_j^n(k+1) = f(v(k), i(k), x_j^n(k), u_j^n(k), k)$$

$$I^n(k+1) = f(v(k), i(k), x(k), u(k), k)$$

$$V^n(k+1) = f(v(k), i(k), x(k), u(k), k)$$

$$YV(k+1) = b(v(k), i(k), I^n(k), V^n(k), k)$$

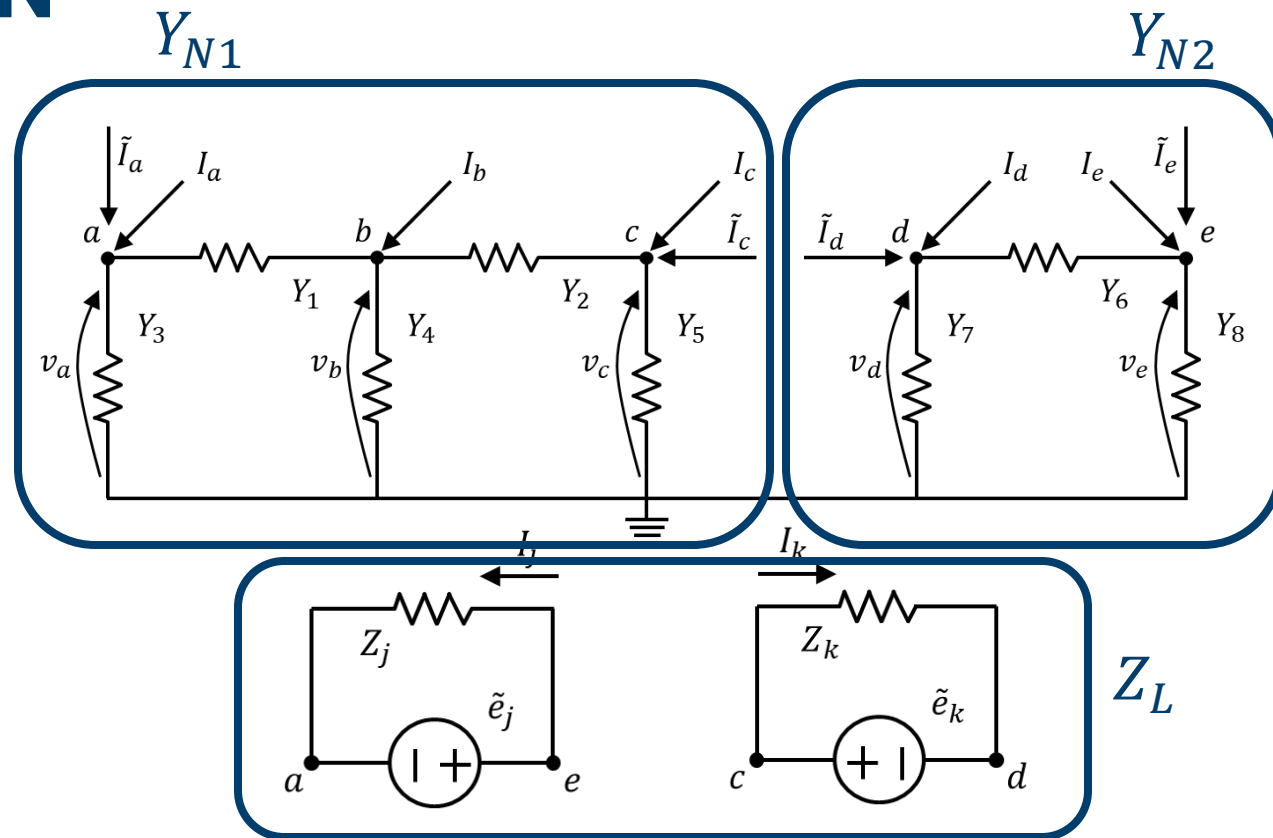
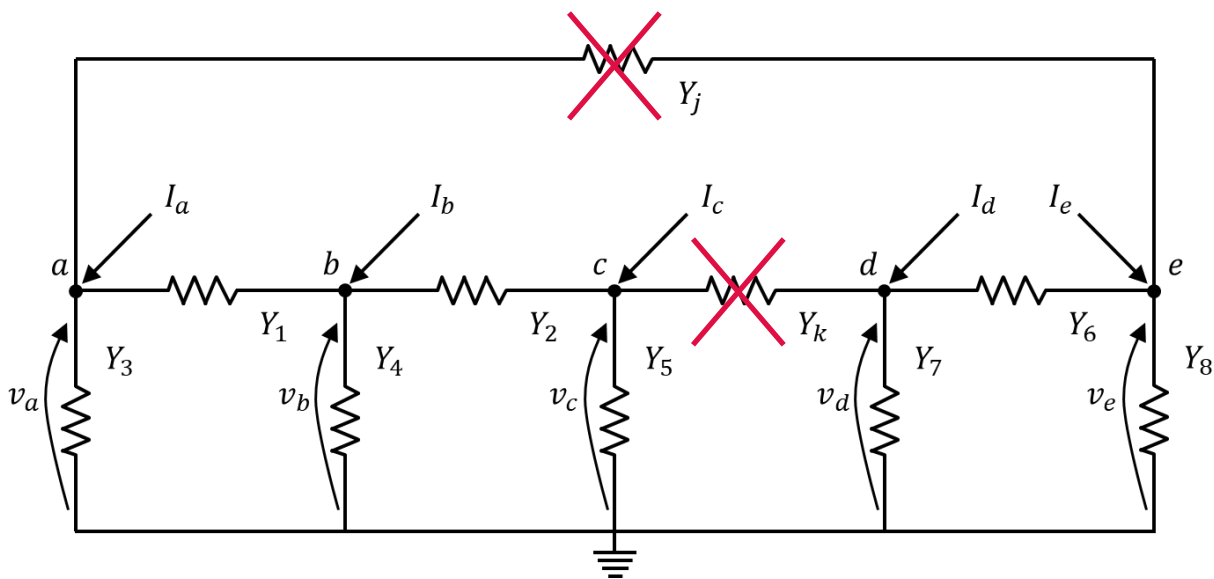
Solved in parallel

Still quite large

POWER SYSTEM SIMULATION

A parallel approach

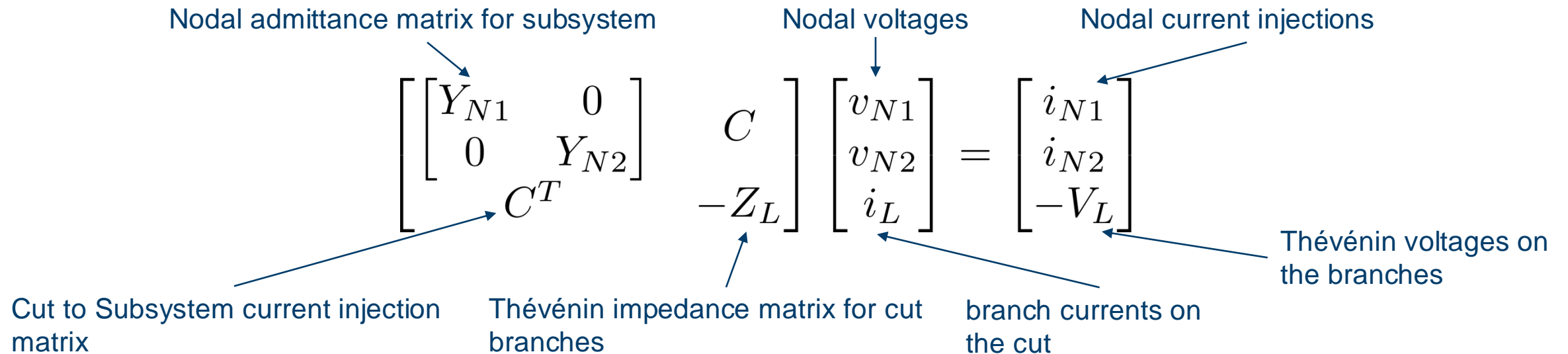
$$YV = b$$



$$\begin{bmatrix} \begin{bmatrix} Y_{N1} & 0 \\ 0 & Y_{N2} \end{bmatrix} \\ C^T \end{bmatrix} \begin{bmatrix} v_{N1} \\ v_{N2} \\ i_L \end{bmatrix} = \begin{bmatrix} i_{N1} \\ i_{N2} \\ -V_L \end{bmatrix}$$

POWER SYSTEM SIMULATION

A parallel approach



POWER SYSTEM SIMULATION

A parallel approach

$$\begin{bmatrix} 1 & 0 & & Y_{N1}^{-1}C_{N1} \\ 0 & 1 & & Y_{N2}^{-1}C_{N2} \\ 0 & 0 & C_{N1}^T Y_{N1}^{-1}C_{N1} + C_{N2}^T Y_{N2}^{-1}C_{N2} + Z_L & \end{bmatrix} \begin{bmatrix} v_{N1} \\ v_{N2} \\ i_L \end{bmatrix} = \begin{bmatrix} Y_{N1}^{-1}h_{N1} \\ Y_{N2}^{-1}h_{N2} \\ C_{N1}^T Y_{N1}^{-1}h_{N1} + C_{N2}^T Y_{N2}^{-1}h_{N2} + V_L \end{bmatrix}$$

➡ Subsystems are independent of each other except of the cut.

Simulation steps in each time step

1. Solve DAE for **components** to obtain injections (in parallel)
2. Solve **Network equations** over the cut
3. Solve **Network equations** of the subsystems (in parallel)

POWER SYSTEM SIMULATION

Optimize the partitioning

- Minimize cut overhead
- Balance the size of the subsystems
- Balance computational load of components between subsystems

Can we formulate the partitioning problem as a QUBO problem ?


GRID PARTITIONING FOR PARALLEL SIMULATION

How to solve QUBOs using Quantum Annealing

Problem Hamiltonian

Quadratic **U**nconstrained **B**inary **O**ptimization:

$$\min_{\{x_n \in [0,1]\}} Q$$
$$Q = \sum_{n,m} x_n Q_{n,m} x_m$$


$$s_n = 2x_n - 1$$

Find ground state of **Ising Hamiltonian**

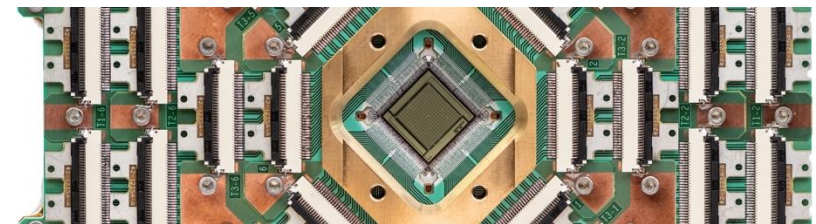
GRID PARTITIONING FOR PARALLEL SIMULATION

How to solve QUBOs using Quantum Annealing

D-Wave Quantum Annealing

- Implementation of „AQC“ to solve optimization problems
- Initial / Driver Hamiltonian H_D is fixed
- Problem Hamiltonian H_P is programmable Ising Hamiltonian
- QPU: currently 5000+ flux qubits and 40000+ couplers arranged in Pegasus topology

➔ Not AQC (not universal) since a open system and limited choice for H_P, H_D



D-Wave Advantage™
System JUPSI (JSC)

GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

- Simulation steps in each time step:
 1. Solve DAE for **components**
 2. Solve **Network eq.** over the **cut**
 3. Solve **Network eq.** of the **subsystems**

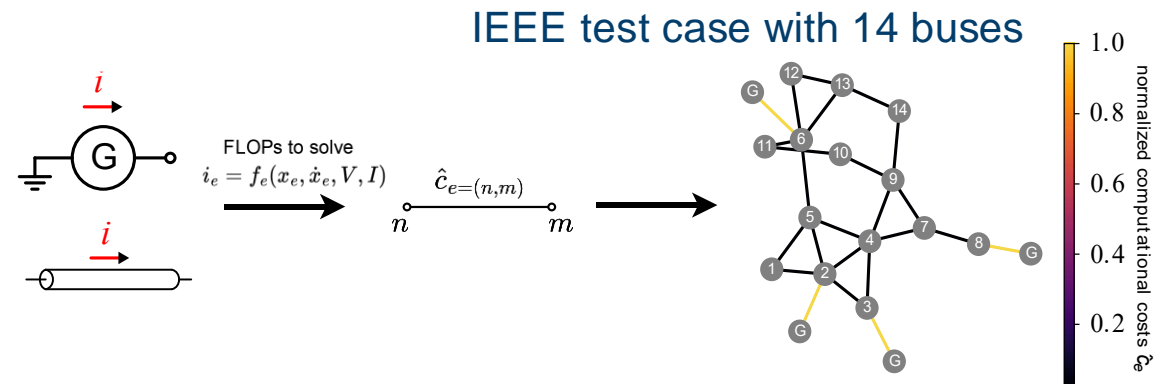


- Objectives for partitioning:
 1. Reduce Idle time components in Subs.
 2. Minimize overhead cut
 3. Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

- Create Graph where every **component** is an **edge** and computational cost as edge weight
- **Binary Encoding** of Partitioning

$$x_n = \begin{cases} 1 & \text{node } n \in \text{Part 1,} \\ 0 & \text{node } n \in \text{Part 2.} \end{cases}$$



GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

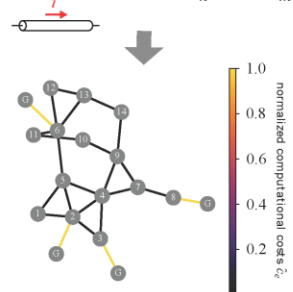
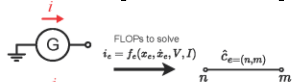
- Simulation steps in each time step:
 1. Solve DAE for **components**
 2. Solve **Network eq.** over the **cut**
 3. Solve **Network eq.** of the **subsystems**



- Objectives for partitioning:
 1. Reduce Idle time components in Subs.
 2. Minimize overhead cut
 3. Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

Graph Representation:



$$x_n = \begin{cases} 1 & n \in \text{Part 1,} \\ 0 & n \in \text{Part 2.} \end{cases}$$

Formulate QUBO objective functions that **estimate** the overhead:

GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

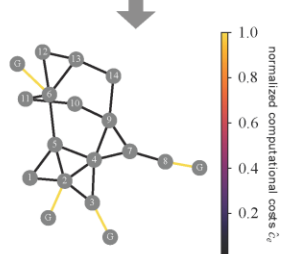
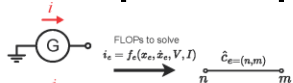
- Simulation steps in each time step:
 - Solve DAE for **components**
 - Solve **Network eq.** over the **cut**
 - Solve **Network eq.** of the **subsystems**



- Objectives for partitioning:
 - Reduce Idle time components in Subs.
 - Minimize overhead cut
 - Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

Graph Representation:



$$x_n = \begin{cases} 1 & n \in \text{Part 1,} \\ 0 & n \in \text{Part 2.} \end{cases}$$

Formulate QUBO objective functions that **estimate** the overhead:

1. Balance components:

$$Q_{comp} = \left(\sum_{e:\text{Part 1}} c_e - \sum_{e:\text{Part 2}} c_e \right)^2 = \left(2 \sum_{n \in V_{bus}} \alpha(n) x_n - \sum_{n \in V_{bus}} \alpha(n) \right)^2$$

$$\alpha(n) := \frac{1}{2} \sum_{m \in V_{bus}} c_{(n,m)} + \sum_{m \in V_{gen}} c_{(n,m)}$$

GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

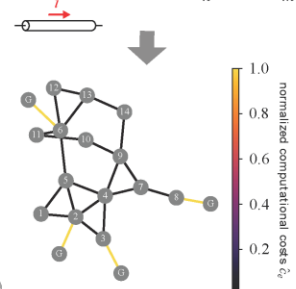
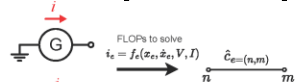
- Simulation steps in each time step:
 1. Solve DAE for **components**
 2. Solve **Network eq.** over the **cut**
 3. Solve **Network eq.** of the **subsystems**



- Objectives for partitioning:
 1. Reduce Idle time components in Subs.
 2. Minimize overhead cut
 3. Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

Graph Representation:



$$x_n = \begin{cases} 1 & n \in \text{Part 1,} \\ 0 & n \in \text{Part 2.} \end{cases}$$

Formulate QUBO objective functions that **estimate** the overhead:

2. Minimize overhead due to cut:

$$Q_{cut} \approx \sum_{e: \text{Cut}} c_e + 4(N-1)M_{cut} = \frac{1}{2} \sum_{n,m \in V_{bus}} A_{n,m} (c_{(n,m)} + 4(N-1)) (-2x_n x_m + x_n + x_m)$$

component cost

Forward-Backward substitution for LU decomposition + updating injections

GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

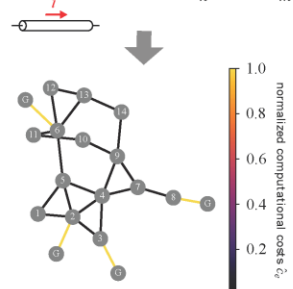
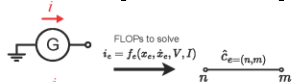
- Simulation steps in each time step:
 1. Solve DAE for **components**
 2. Solve **Network eq.** over the **cut**
 3. Solve **Network eq.** of the **subsystems**



- Objectives for partitioning:
 1. Reduce Idle time components in Subs.
 2. Minimize overhead cut
 3. Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

Graph Representation:



$$x_n = \begin{cases} 1 & n \in \text{Part 1,} \\ 0 & n \in \text{Part 2.} \end{cases}$$

Formulate QUBO objective functions that **estimate** the overhead:

3. Balance Subnetwork size:

$$Q_{network} = (2N - 1)^2 \left(2 \sum_n x_n - N \right)^2$$

Difference for Forward-Backward substitution based on LU decomposition

GRID PARTITIONING FOR PARALLEL SIMULATION

QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids

Parallel Simulation of Partitioned Grid

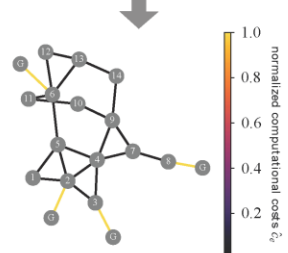
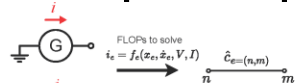
- Simulation steps in each time step:
 1. Solve DAE for **components**
 2. Solve **Network eq.** over the **cut**
 3. Solve **Network eq.** of the **subsystems**



- Objectives for partitioning:
 1. Reduce Idle time components in Subs.
 2. Minimize overhead cut
 3. Reduce Idle time subnetworks

QUBO formulation for optimal partitioning

Graph Representation:



$$x_n = \begin{cases} 1 & n \in \text{Part 1,} \\ 0 & n \in \text{Part 2.} \end{cases}$$

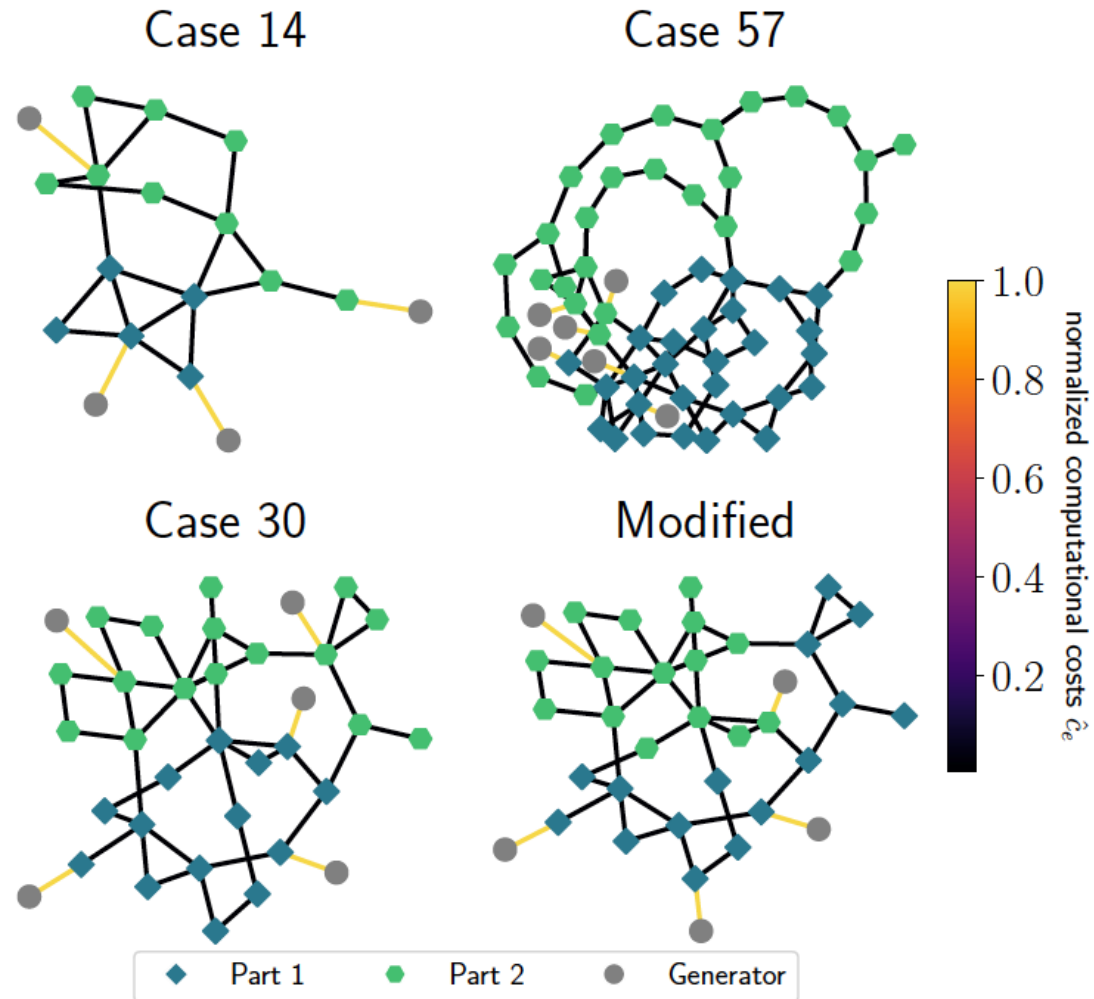
Full QUBO:

$$Q = Q_{comp} + Q_{cut} + Q_{net}$$

Estimates the complete overhead of a partitioning.

GRID PARTITIONING FOR PARALLEL SIMULATION

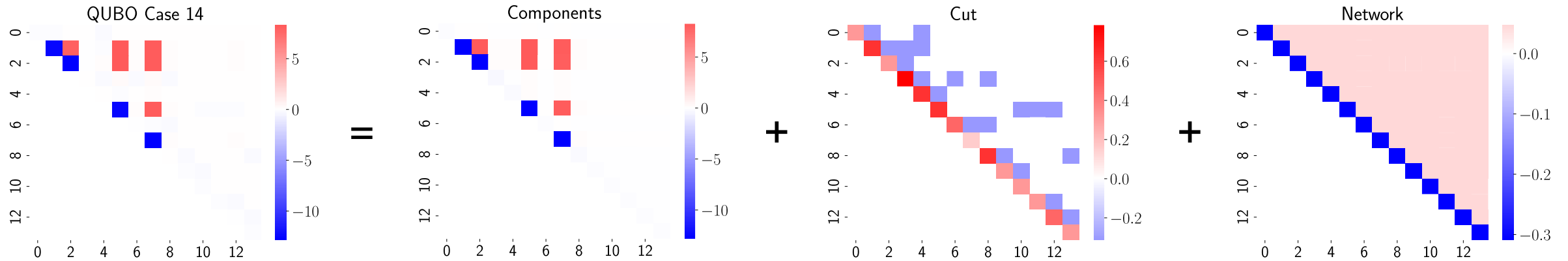
QUBO for Optimal Graph Partitioning for Parallel Simulation of Power Grids – Gurobi solution



GRID PARTITIONING FOR PARALLEL SIMULATION

Solving on D-Wave annealer

The Need For Embedding

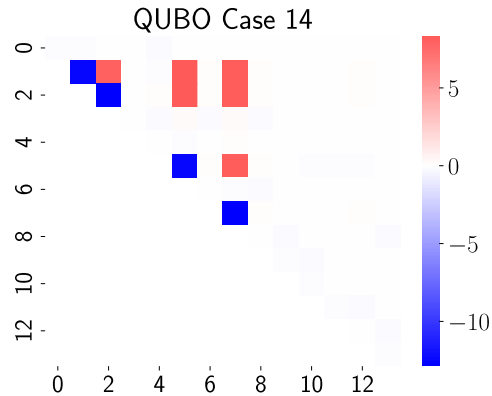


- Some QUBO terms are fully connected

GRID PARTITIONING FOR PARALLEL SIMULATION

Solving on D-Wave annealer

The Need For Embedding



- Some QUBO terms are fully connected
- QPU has limited qubit connectivity

➔ Need (minor) Embedding: **chain** of physical qubit represent one logical qubit

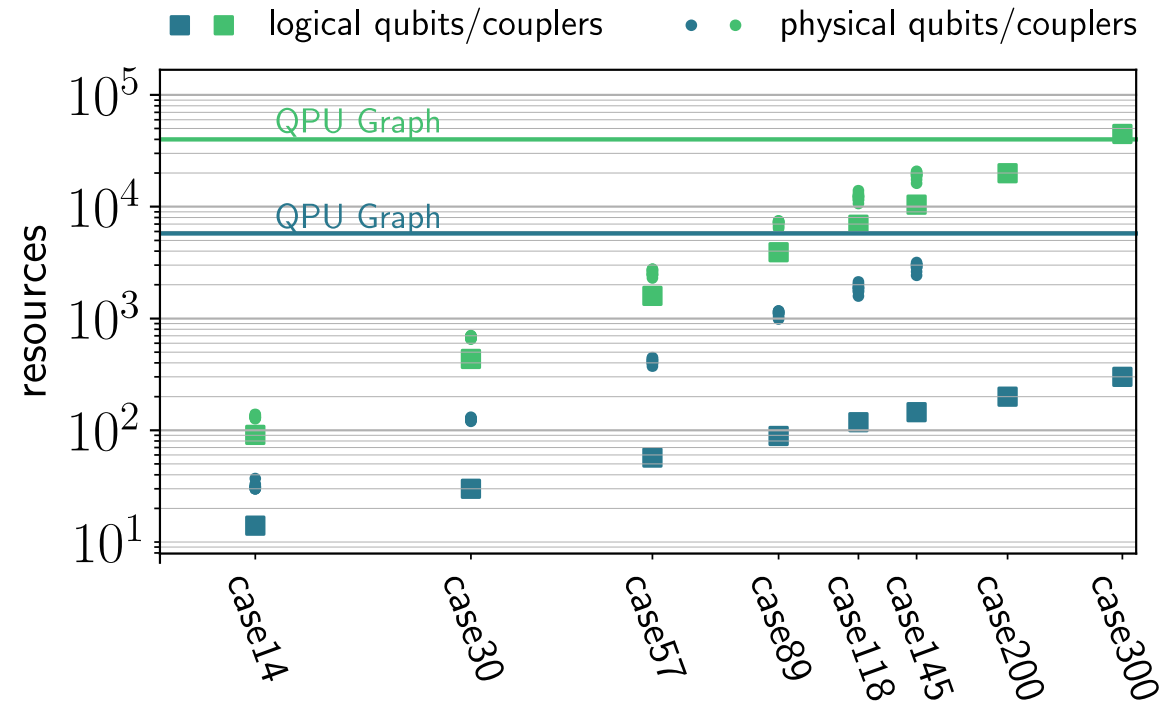
GRID PARTITIONING FOR PARALLEL SIMULATION

Solving on D-Wave annealer

The Need For Embedding

Limited connectivity on QPU \longrightarrow Dense QUBOs must be embedded

Consequences of Embedding



GRID PARTITIONING FOR PARALLEL SIMULATION

Solving on D-Wave annealer

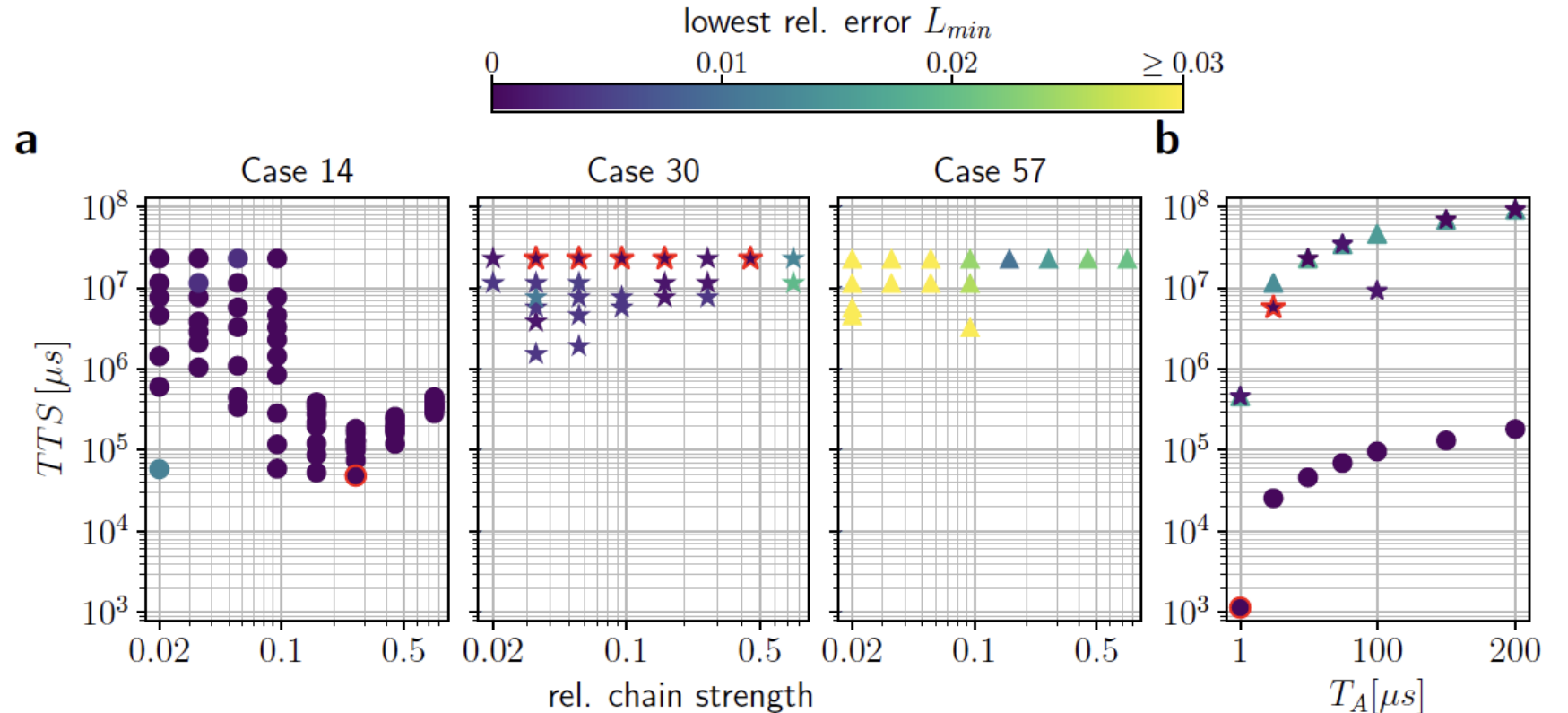
Relative Error

$$L = \frac{|E_{min, global} - E_{sampled}|}{|E_{min, global}|}$$

$$L_{min} := \min_{E_{sampled}} L.$$

Time to Solution

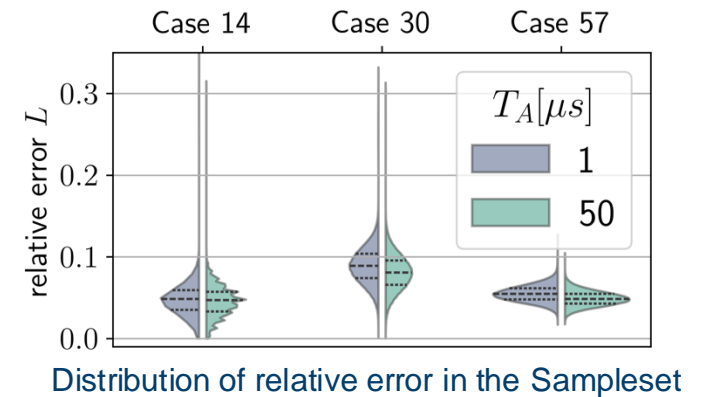
$$TTS(p_S, T_A) := \frac{\log(1 - p_S)}{\log(1 - \frac{N_{opt}}{N_s})} T_A,$$



GRID PARTITIONING FOR PARALLEL SIMULATION

Outlook: What could be next for partitioning?

1. Embeddings are the major bottleneck for scalability:
 - Prune the QUBO, so it becomes more sparse
 - Test other reduction methods (Ohzeki. et al: „Breaking limitation of quantum annealer in solving optimization problems under constraints“)
2. Near-optimal solutions can be sampled fast:
 - Evaluate near optimal solutions in the simulation loop
 - What is the optimal number of samples to achieve fast simulation with high probability.
3. The QUBO does not enforce that the subsystems are connected:
 - Try to formulate additional constraints that are efficient.



CONCLUSION

- A preprint of the paper used for this lecture can be found at:
<https://arxiv.org/pdf/2408.04097>
- Quantum computing can support power system simulation and optimization but:
 - What is the gain ?
 - Which problem ?
- If you are interested in working on the topic contact me!
<https://www.fz-juelich.de/en/ice/ice-1>
Andrea Benigni a.benigni@fz-juelich.de