

# DLR QUANTUM COMPUTING INITIATIVE

Quelle: planqc

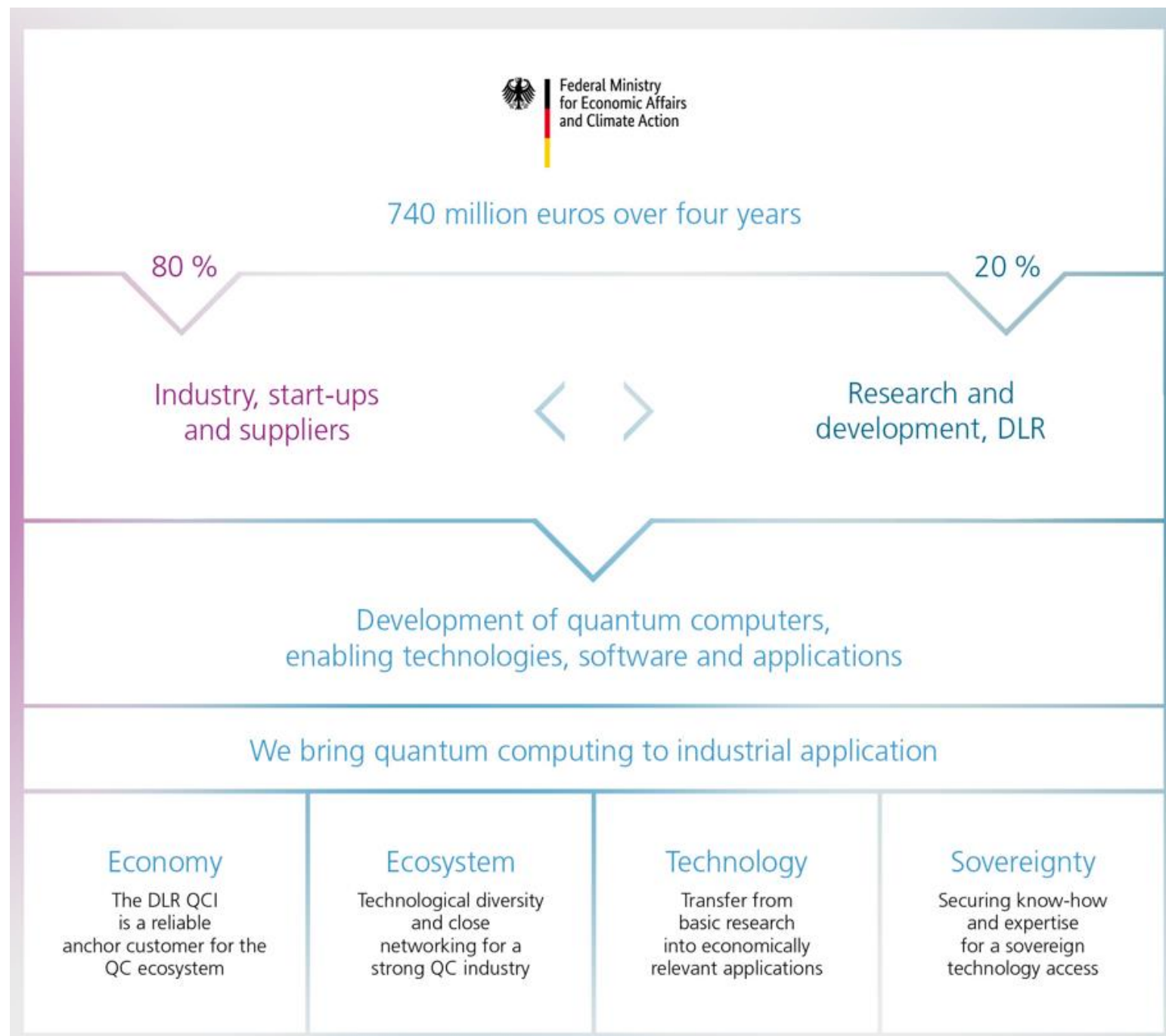


## DLR Quantum Computing Initiative

*We shape the  
quantum computing ecosystem*

- Bundled technology, expertise & infrastructure
- Research and development on the basis of DLR areas of competence
- Procurement of complete QC systems and components through open-technology competitive tenders
- Upgrading by commissioning the necessary basic technologies and applications

[qci.dlr.de](https://qci.dlr.de)





# DLR Quantum Computing Initiative

*We shape the  
quantum computing ecosystem*

1

Unique  
ecosystem

2

Innovation  
centers

19

DLR  
Institutes

22

Industry  
partners

6

Qubit  
technologies

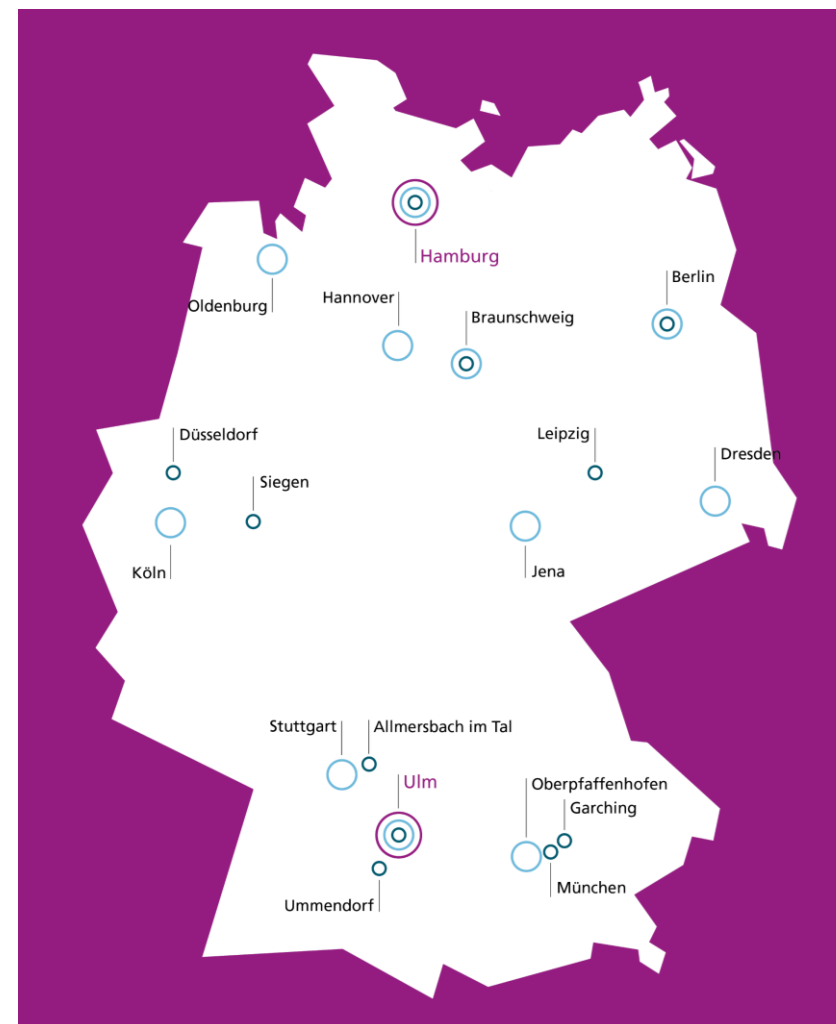
17

Hardware  
projects

22

SW / App  
projects

-  Innovationszentren
-  DLR-Institute
-  Industriepartner



# QCI – Industry Partners



# Technologies



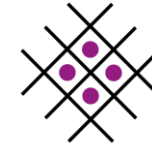
There are many ways to realize qubits and thus quantum computers in hardware.  
That is why we commission the development of different hardware platforms.



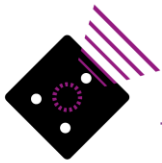
Analogue Computers



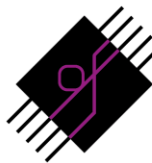
Ion traps



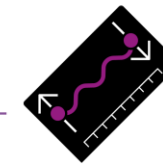
Neutral atoms



NV Centers



Photons



Spin enabling

→ What is the best approach to a universal, error-corrected quantum computer?





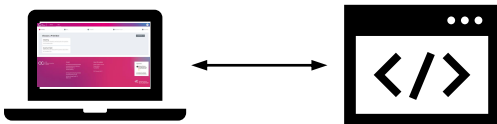
# THE DLR QC ECOSYSTEM

# Goal: Fully Integrative QC Hardware Software Stack

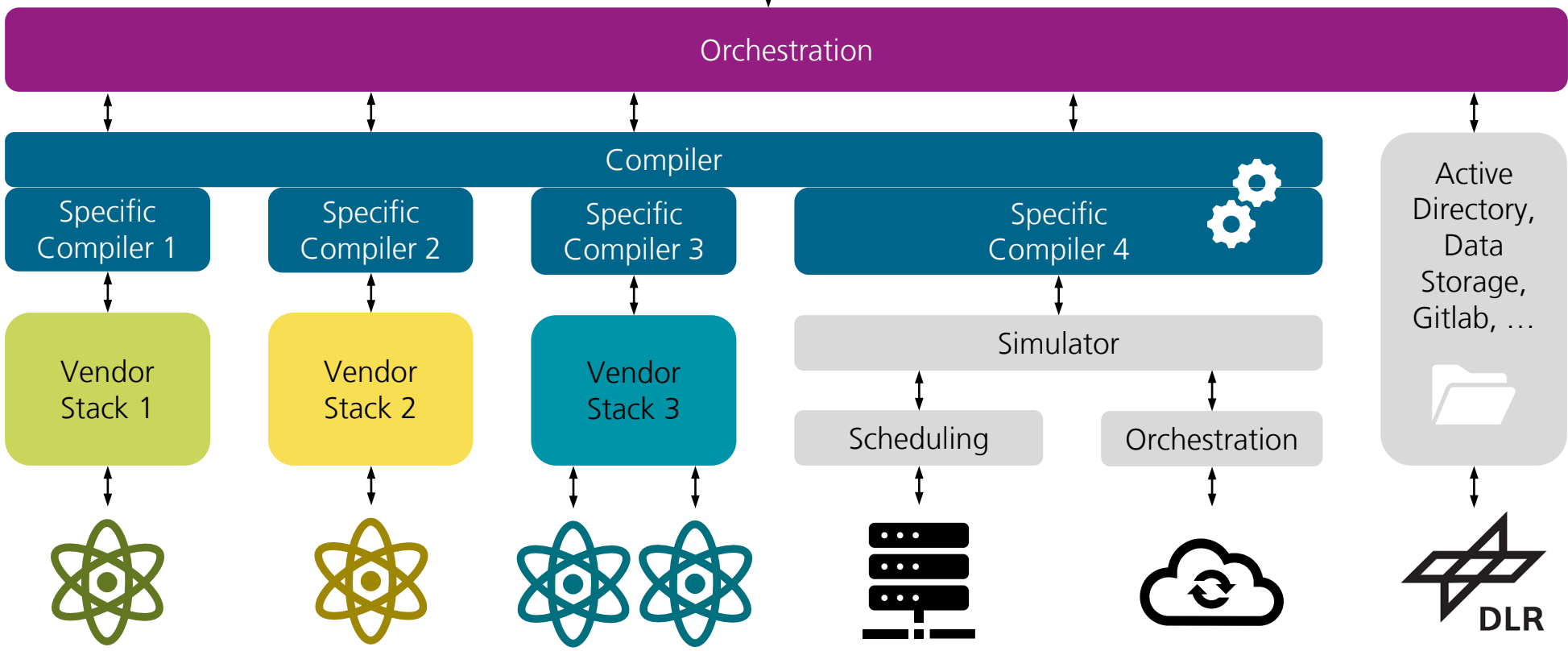


ALQU

planqc d-fine

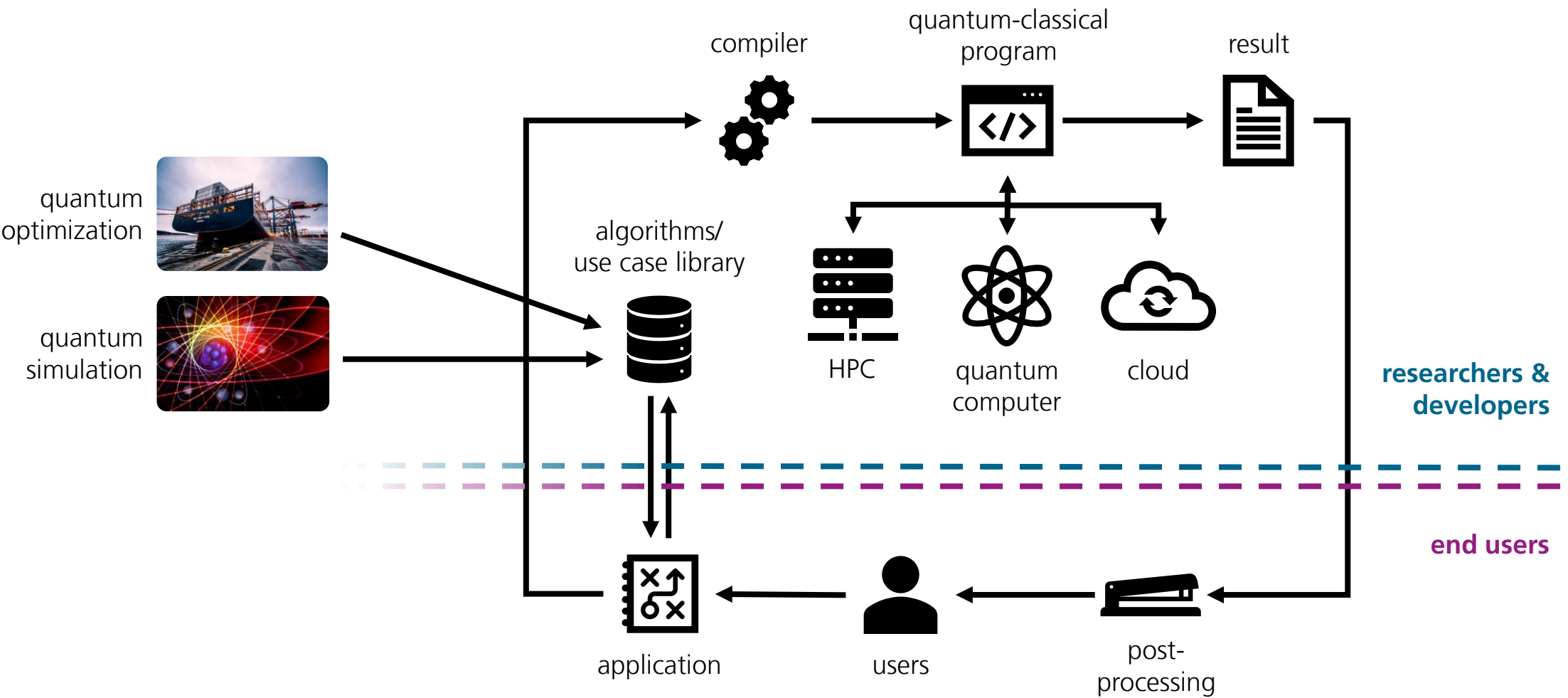


Quantum Software – Aspects of Theory and System Design  
Basermann et al., 2024, Carbonelli et al., 2024



→ QC hardware is only a very small part of the full system

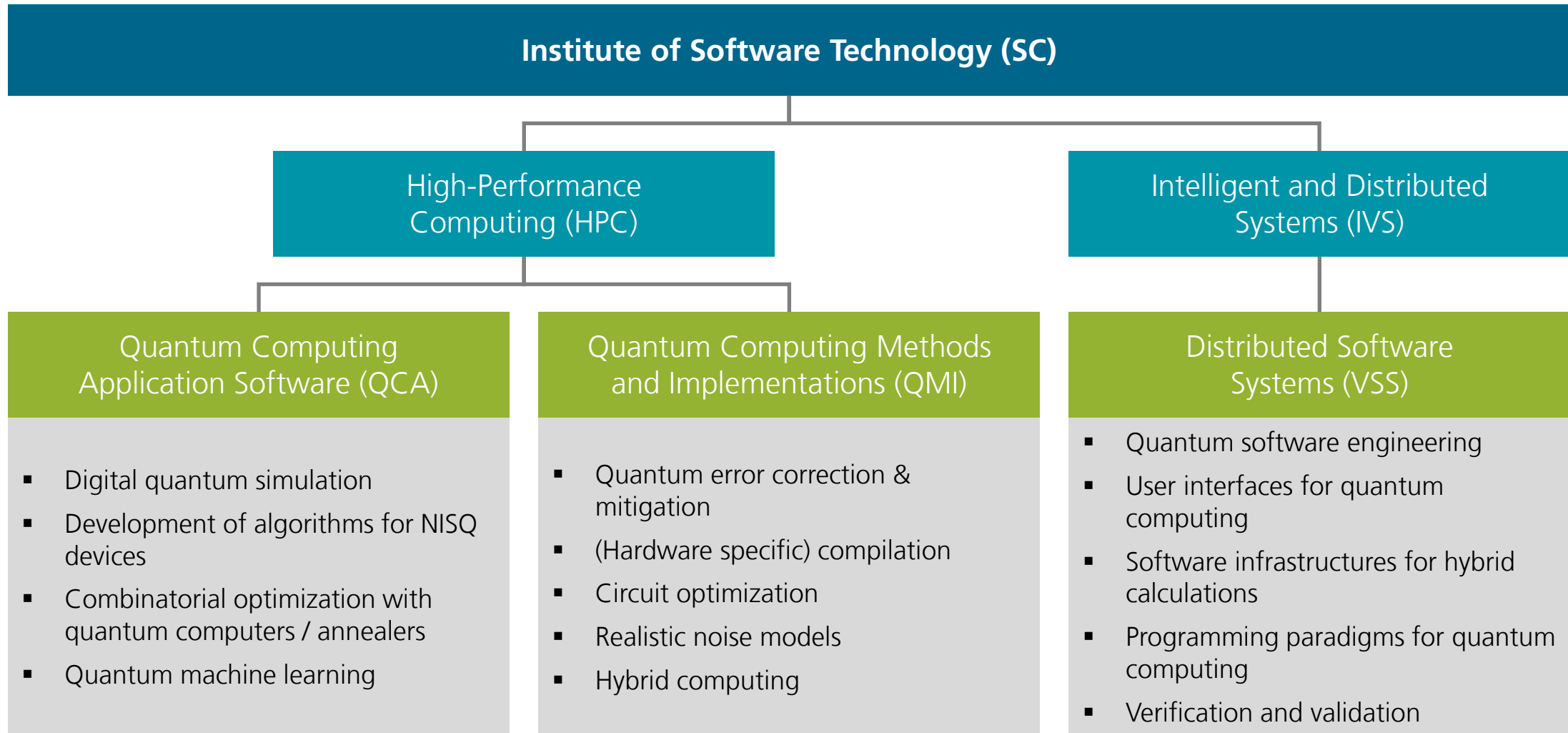
# Ideal Workflow of Using Quantum Computers



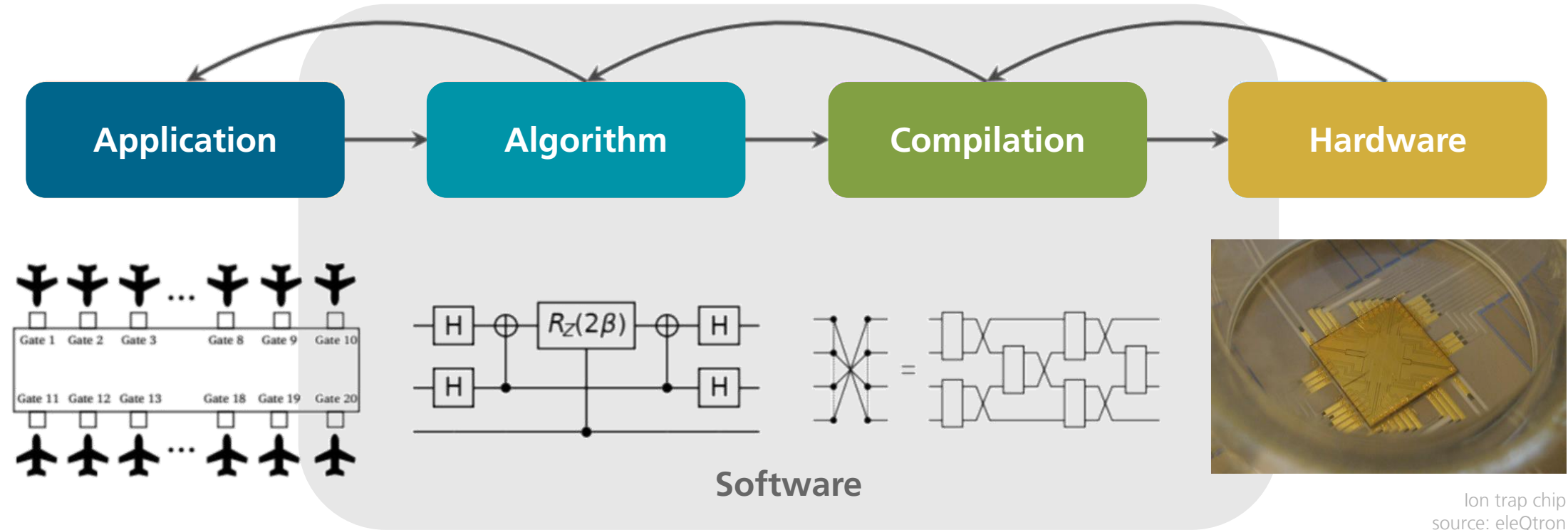


# QC @ DLR-SC

# Groups Working on QC at DLR-SC

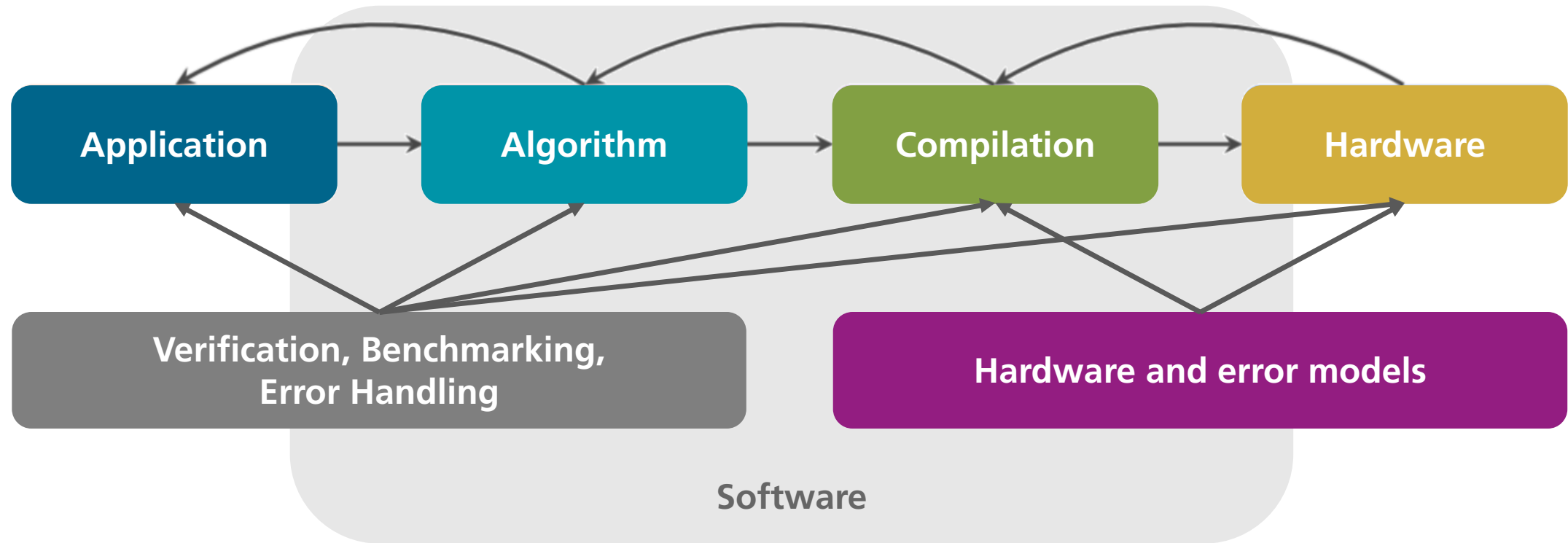


# Hardware-Software Co-Design



- ➔ **Identification** of promising QC applications on the one hand as well as QC platforms (digital, analog, topological) on the other hand
- ➔ **Development** of new algorithms and application software for NISQ devices with the goal to find quantum advantage by optimally exploiting hardware specifics

# Hardware-Software Co-Design



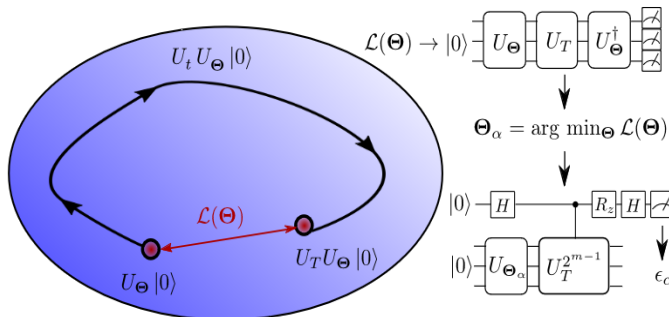
- Access and therefore adaptability at all levels of the QC stack
- Suitable abstraction layers and solution patterns, in particular enabling hybrid computing
- Integration of SW/HW manufacturers and users
- Comparison of DLR QC hardware with state-of-the-art classical methods



# Quantum Applications

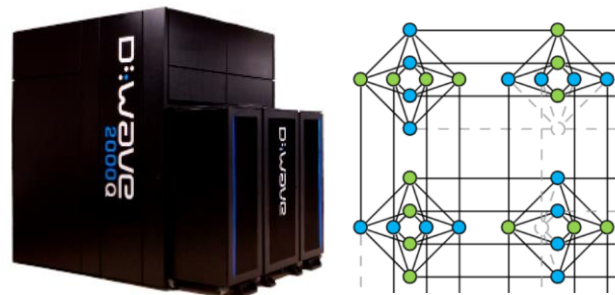
## Digital Quantum Simulation

- New algorithms for quantum many body dynamics
- Development of quantum simulation software for NISQ devices



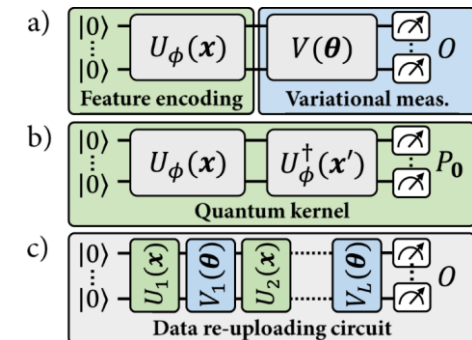
## Combinatorial Optimization

- Combinatorial optimization problems around quantum annealers
- Development of open-source software library (quark and Co.)



## Other Quantum Applications and Quantum-inspired Algorithms

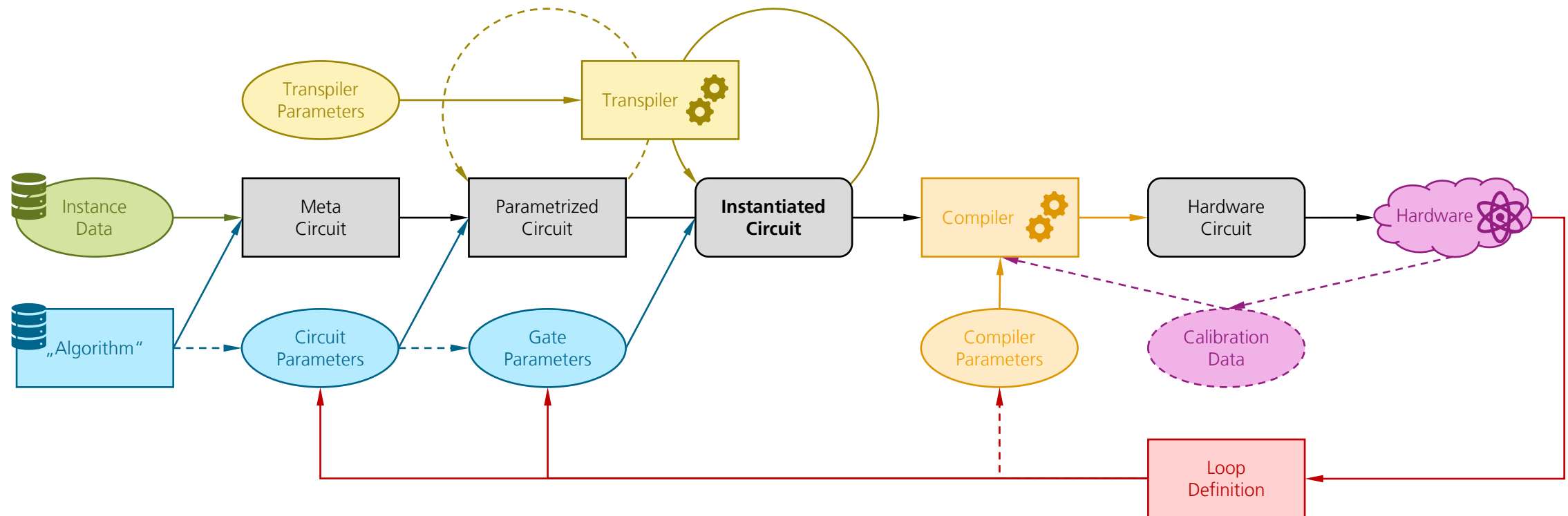
- Quantum machine learning
- Description of new quantum computing platforms (e.g. topological QCs)
- State-of-the-art numerical software for QC benchmarks





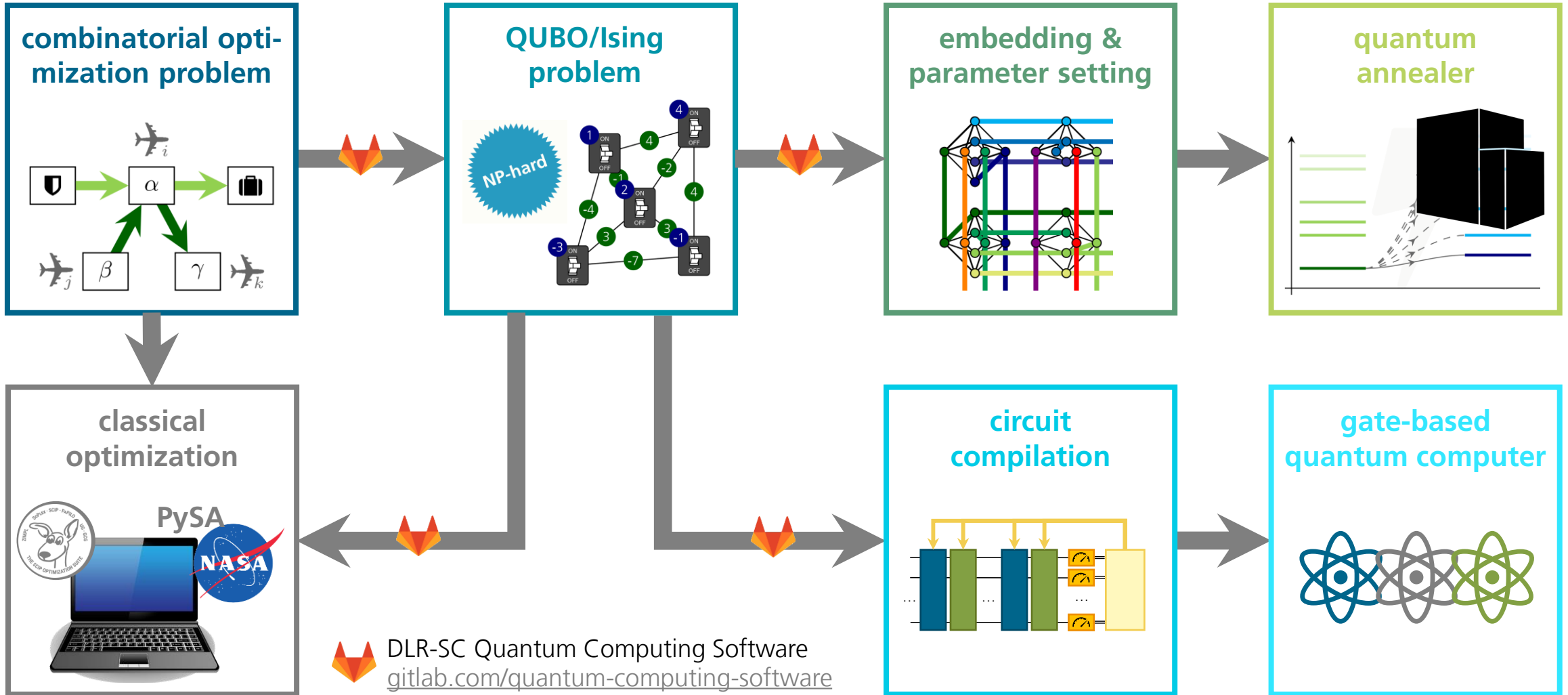
# Support of Hybrid Quantum Algorithms in QC Ecosystem

- Problem: input output behaviour and definition of the interaction with classical systems
- Representation of loops with regard of scheduling and orchestration



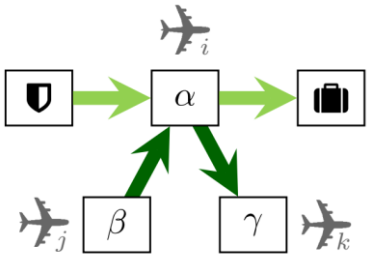
# QUANTUM OPTIMIZATION SOFTWARE

# Combinatorial Optimization QC Software Stack



# Combinatorial Optimization Problems

## combinatorial optimization problem

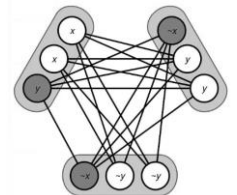
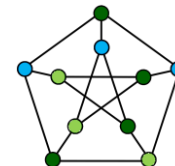
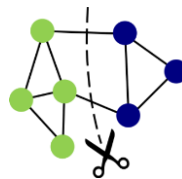


instance  
database

[10.5281/zenodo.13944133](https://zenodo.org/record/13944133)

## Implemented Problems

- Minimum k-Union
- Graph Partitioning
- Subset Sum
- Knapsack
- Max Cut
- Prime Factorization
- Random/Arbitrary Ising Model
- Maximum Colorable Subgraph
- Flight-Gate Assignment
- Traveling Salesperson



quapps

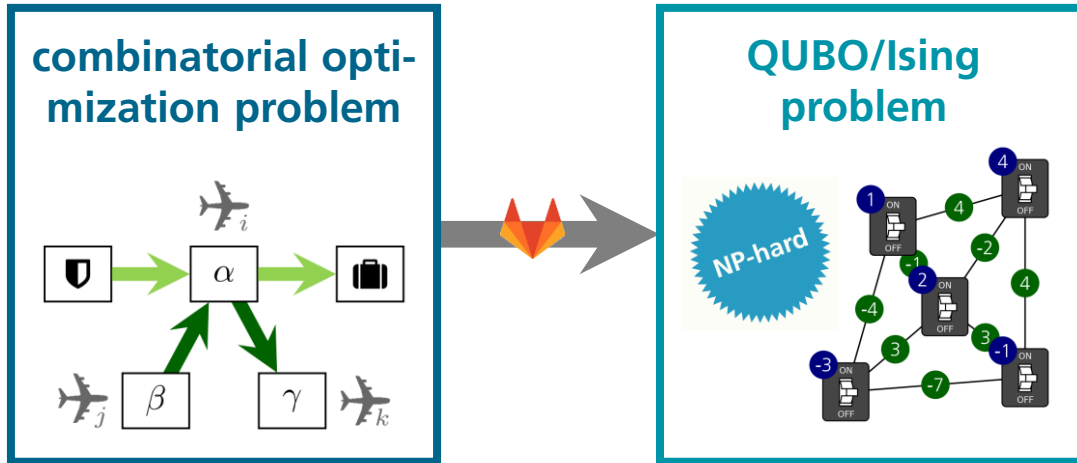


<https://anaconda.org/dlr-sc/quapps>  
conda install dlr-sc::quapps

## Planned

- Autoclave Loading
- Transmission Expansion Planning
- Antenna Alignment
- Satisfiability
- ...

# Transformation to Ising Problems



$$\begin{aligned} &\text{minimize } p(x) \\ &\text{subject to } \ell_1 \leq q_1(x) \leq u_1, \\ &\quad \dots, \\ &\quad \ell_m \leq q_m(x) \leq u_m, \\ &\quad x \in \mathbb{Z}^n \end{aligned}$$

$$\begin{aligned} &\text{minimize } \sum_{i=1}^N W_i x_i + \sum_{i < j=1}^N S_{ij} x_i x_j \\ &\text{subject to } x \in \{0, 1\}^N / \{-1, 1\}^N \end{aligned}$$



<https://anaconda.org/dlr-sc/quark>  
`conda install dlr-sc::quark`



[10.18420/inf2023\\_123](https://doi.org/10.18420/inf2023_123)  
[10.5281/zenodo.13846762](https://doi.org/10.5281/zenodo.13846762)

## Main Transformation Steps:

- Encoding of the variables
- Reformulation of constraints to penalty terms added to the objective
- Choosing penalty weights
- Reduction of higher-degree polynomials



# Reduction of Higher-degree Polynomials

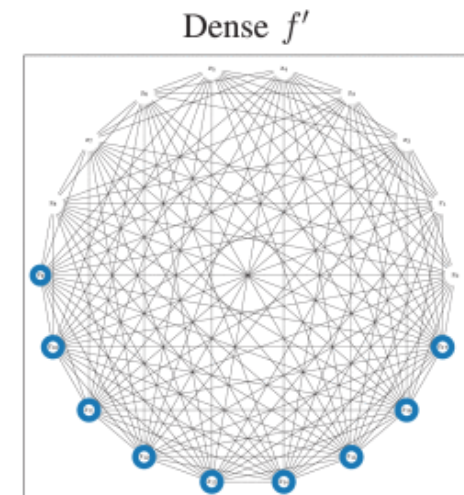
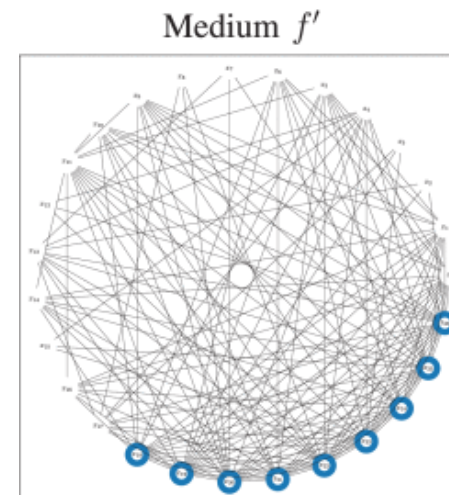
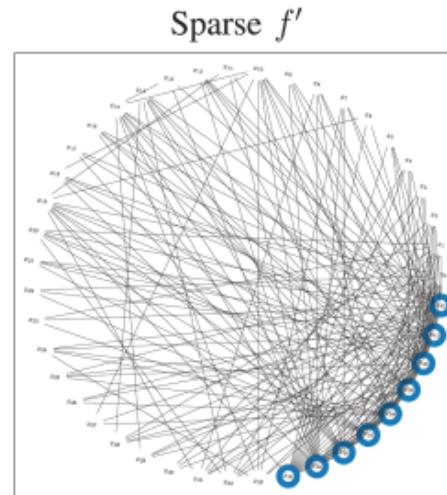
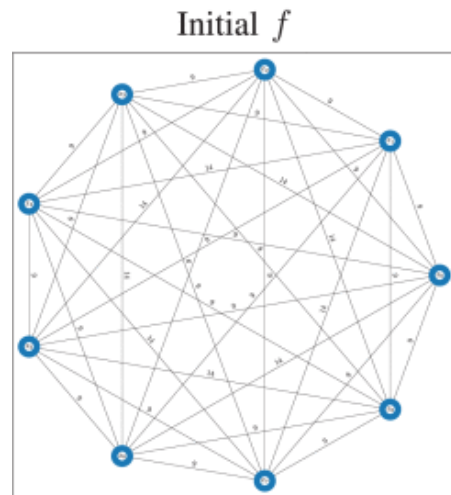


SIEMENS



10.1109/QSW62656.2024.00018

- By Lukas Schmidbauer, Elisabeth Lobe, Karen Wintersperger, Wolfgang Maurer
- Analyzed the influence of the reduction methods implemented in **quark** on the resulting Ising interaction graph
- And further on **their Impact on QAOA Circuits**



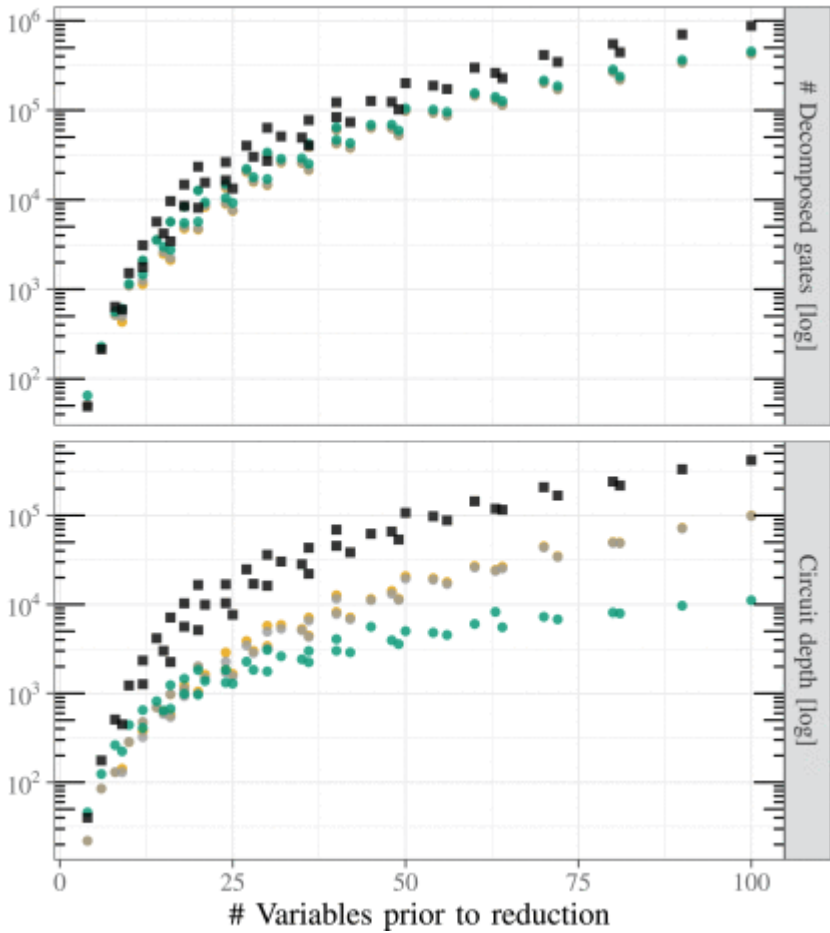
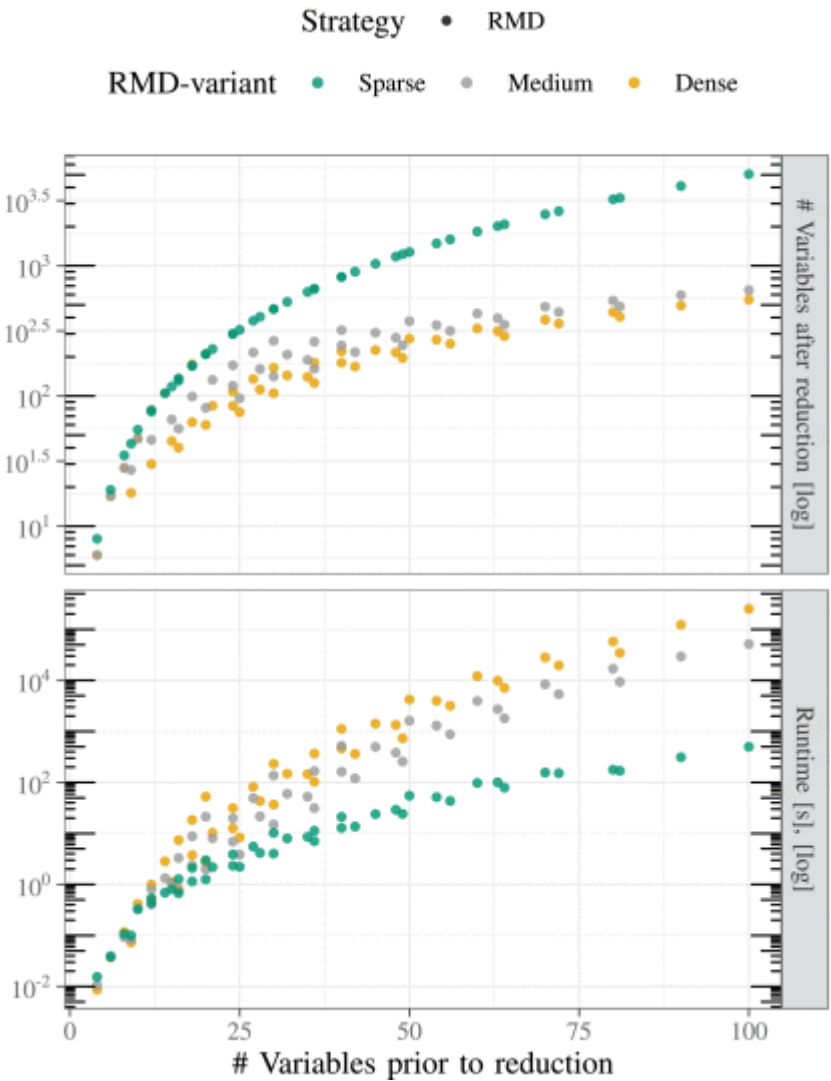
# Reduction Impact on QAOA Circuits



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 10.1109/QSW62656.2024.00018



# Faster Reduction of Higher-degree Polynomials

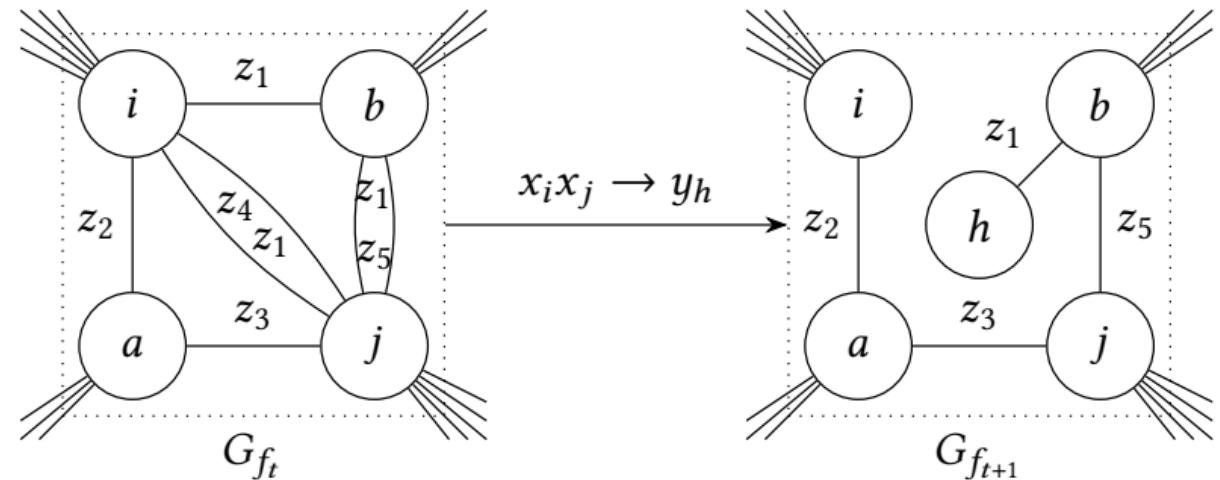
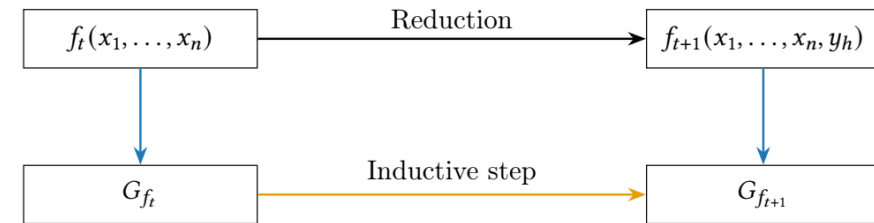


[10.48550/arXiv.2411.19934](https://arxiv.org/abs/10.48550/arXiv.2411.19934)

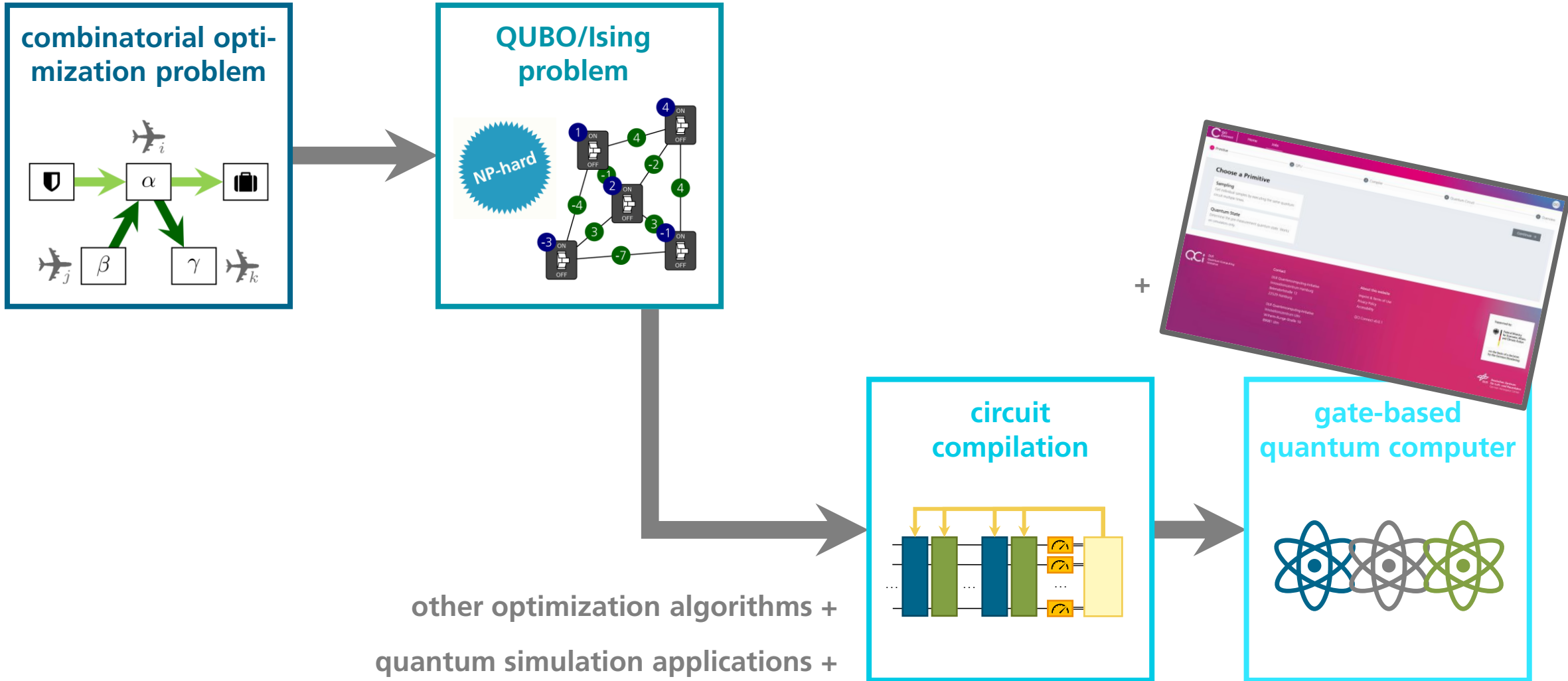


[https://github.com/lfd/Fast\\_Quadratisation](https://github.com/lfd/Fast_Quadratisation)

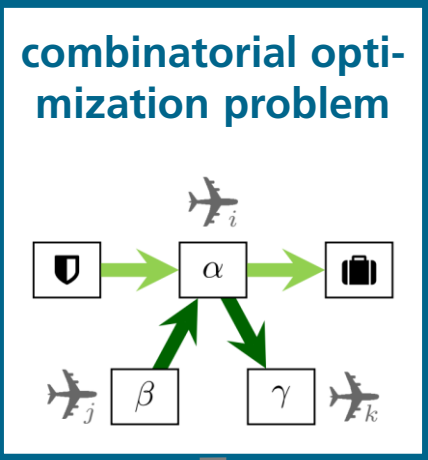
- By Lukas Schmidbauer, Elisabeth Lobe, Ina Schaefer, Wolfgang Mauerer
- Represented polynomials with a specific multigraph representation
- Based on that a new, **faster reduction** algorithm was developed
- Inductive step updates graph structure
- Rather than scanning the polynomial several times



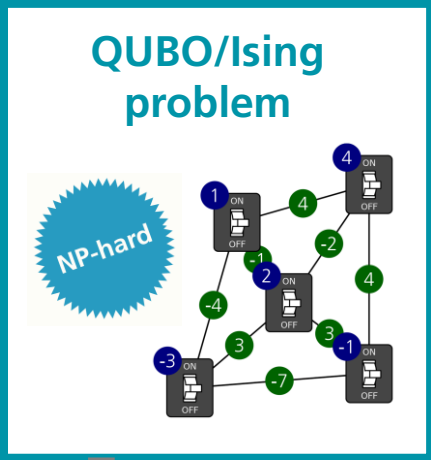
# Use Cases Transformation for QAOA



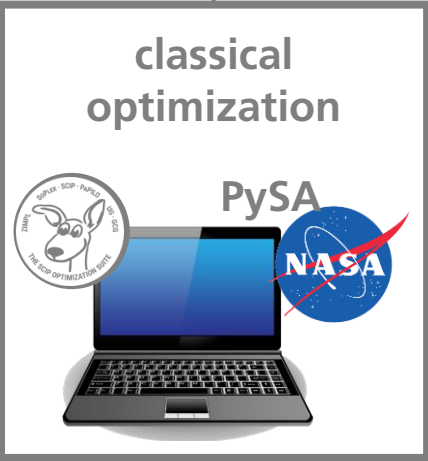
# Classical Solving



+ ...



+ cvxpy solver (DLR-VE)  
+ Benders' Decomposition

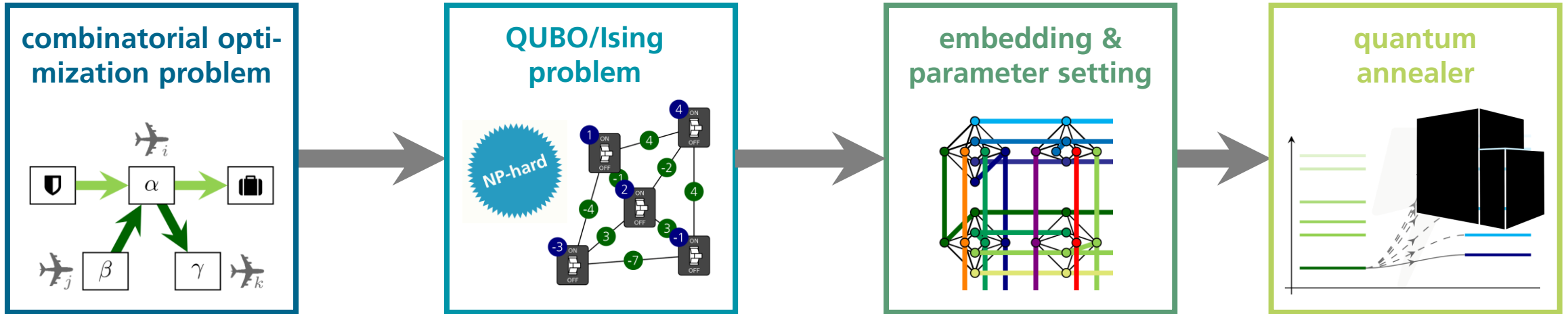


+ McSparse/SpinGlass solver  
+ Selby algorithm

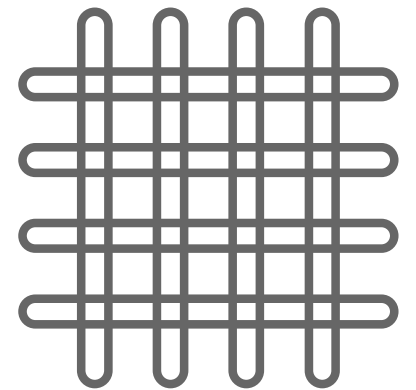




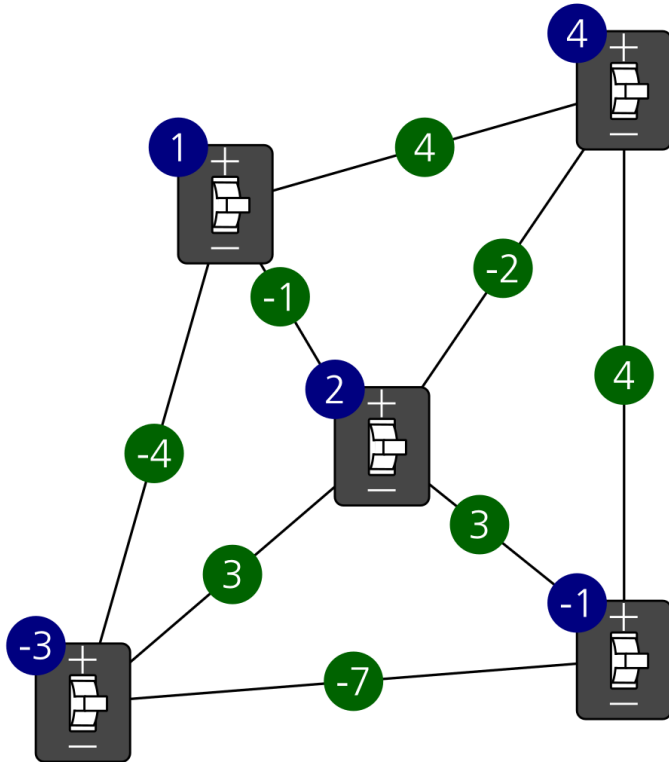
# Use Cases Transformation for Quantum Annealing



- Restricted hardware graph
- "Broken" qubits
- Limited parameter precision



# THE EMBEDDED ISING PROBLEM



## Definition

An **Ising model** over graph  $G$  with **weights**  $W \in \mathbb{R}^{V(G)}$  and **strengths**  $S \in \mathbb{R}_{\neq 0}^{E(G)}$  is a function  $I_{W,S} : \{-1, 1\}^{V(G)} \rightarrow \mathbb{R}$  with

$$I_{W,S}(s) := \sum_{v \in V(G)} W_v s_v + \sum_{vw \in E(G)} S_{vw} s_v s_w.$$

We call  $G$  the **interaction graph** of the Ising model.

## ISING PROBLEM

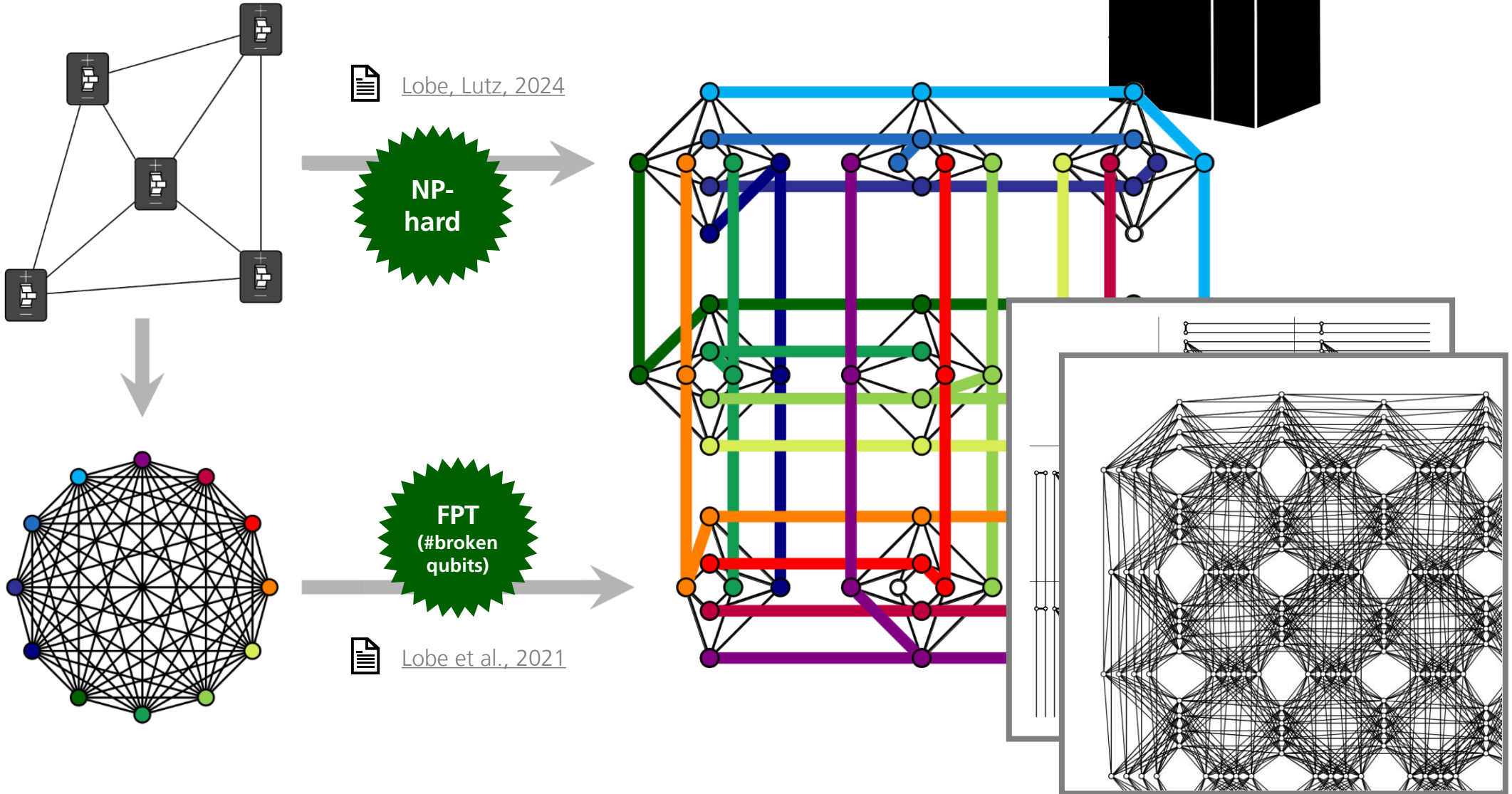
Given a graph  $G$ ,  $W \in \mathbb{R}^{V(G)}$  and  $S \in \mathbb{R}^{E(G)}$ , find  $s$  that solves

$$\min_{s \in \{-1, 1\}^{V(G)}} I_{W,S}(s).$$

NP-hard



# Embedding Problem for D-Wave Hardware



# Embedded Ising Problem

 Lobe, Kaibel, 2023

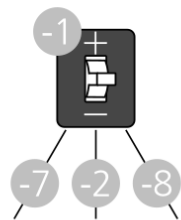


$$\begin{aligned}\bar{I}_{\bar{W}, \bar{S}}(s) &:= \sum_{q \in V(H_\varphi)} \bar{W}_q s_q + \sum_{pq \in E_\varphi \cup E_\delta} \bar{S}_{pq} s_p s_q \\ &= \sum_{v \in V(G)} \left( \sum_{q \in \varphi_v} \bar{W}_q s_q + \sum_{pq \in E(H[\varphi_v])} \bar{S}_{pq} s_p s_q \right) + \sum_{vw \in E(G)} \left( \sum_{pq \in \delta_{vw}} \bar{S}_{pq} s_p s_q \right)\end{aligned}$$

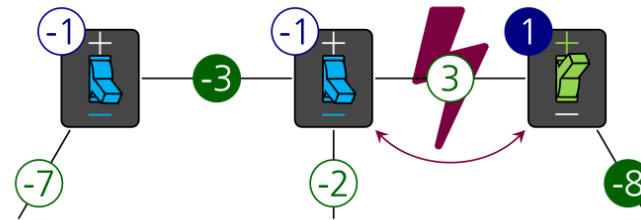
$\swarrow \nearrow W_v t_v$ 
 $\swarrow \nearrow S_{vw} t_v t_w$

- with **proven equivalence** to original problem
- such that (at least optimal) solutions correspond to each other
- based on **synchronization** of the embedded variables

original vertex

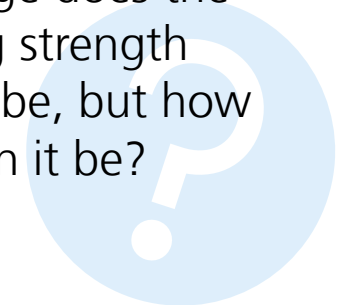


embedded vertices



$$= -1 - 0 - 17 = -18 \text{ ⚡}$$

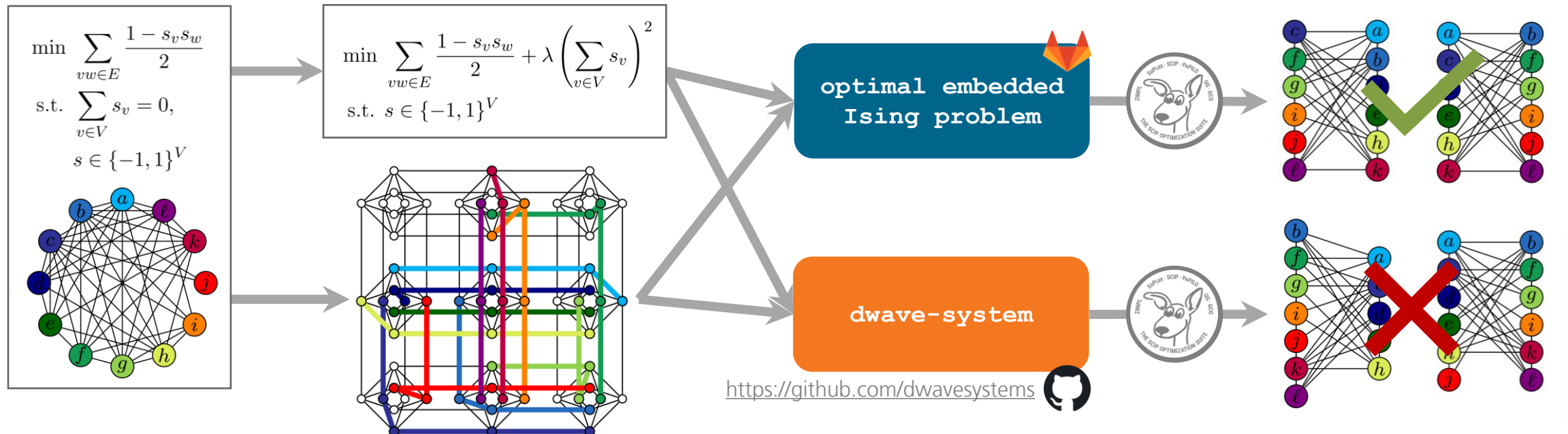
How large does the coupling strength need to be, but how small can it be?





# Get Embedded Ising via “Uniform Torque Compensation”

- Current implementation in D-Wave dwave-ocean-sdk with method `embed_ising`
  - calls `dwave-system.dwave.embedding.chain_strength.uniform_torque_compensation`
- Found **counterexample**
  - where the method does **not** provide an equivalent embedded Ising problem
  - based on the Graph Partitioning Problem with 12 nodes
  - embedded by D-Wave using `minorminer.find_embedding`



# Get the Optimal Embedded Ising Problem



Lobe, Kaibel, 2023



Requires to solve for each individual original vertex:

## GAPPED WEIGHT DISTRIBUTION PROBLEM

Given graph  $G = (V, E)$ ,  $\sigma \in \mathbb{R}_{\geq 0}^V$ ,  $W \in \mathbb{R}_{\geq 0}$  with  $W < \sigma(V)$  and  $\gamma \in \mathbb{R}_{>0}$ :

$\min \vartheta$

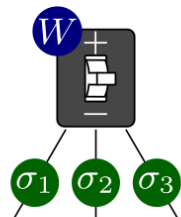
s.t.  $\vartheta \in \mathbb{R}$ ,  $\omega \in \mathbb{R}^V$

$$\vartheta \geq \frac{\min \{ \sigma(S) + \omega(S), \sigma(V \setminus S) - \omega(V \setminus S) \} + \gamma}{|\delta(S)|} \quad \forall \emptyset \neq S \subsetneq V,$$

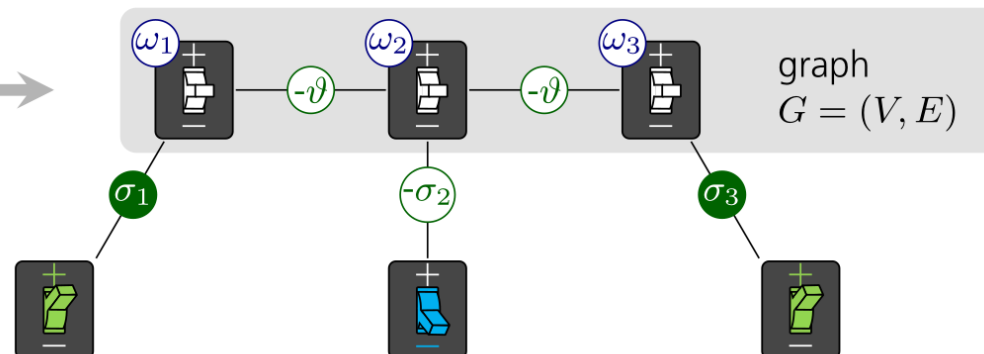
$$\omega(V) = W$$

polynomial  
for trees

original vertex

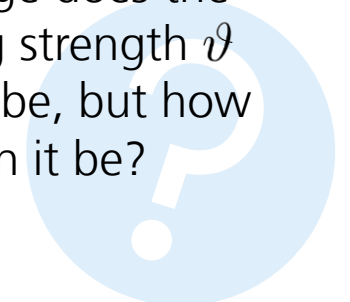


embedded vertices



- cut constraints are redundant for  $S$  or  $V \setminus S$  being not connected
- efficiently solvable** in practical embedding setup

How large does the coupling strength  $\vartheta$  need to be, but how small can it be?



# Get the Optimal Embedded Ising Problem



Lobe, Kaibel, 2023



Lobe, PhD Thesis, 2022



Requires to solve for each individual original vertex:

## GAPPED INTEGER WEIGHT DISTRIBUTION PROBLEM

Given graph  $G = (V, E)$ ,  $\sigma \in \mathbb{N}^V$ ,  $W \in \mathbb{N}$  with  $W < \sigma(V)$  and  $\gamma \in \mathbb{N}_+$ :

min  $\vartheta$

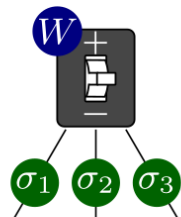
s.t.  $\vartheta \in \mathbb{Z}$ ,  $\omega \in \mathbb{Z}^V$ ,

$$\vartheta \geq \frac{\min \{ \sigma(S) + \omega(S), \sigma(V \setminus S) - \omega(V \setminus S) \} + \gamma}{|\delta(S)|} \quad \forall \emptyset \neq S \subsetneq V,$$

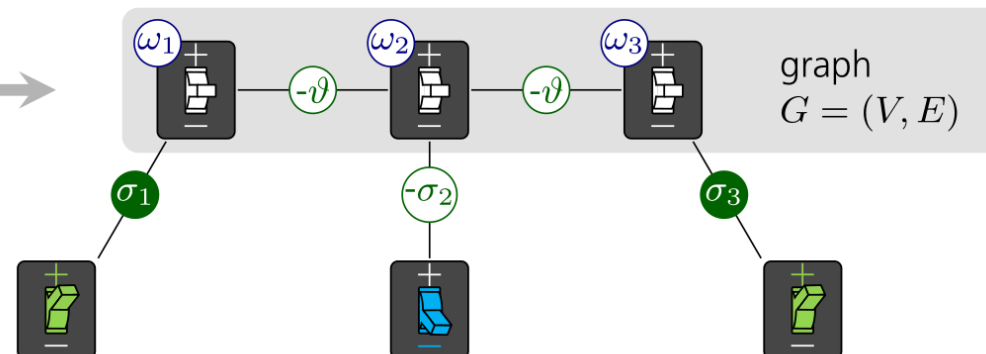
$$\omega(V) = W$$

polynomial  
for trees

original vertex



embedded vertices



- cut constraints are redundant for  $S$  or  $V \setminus S$  being not connected
- efficiently solvable** in practical embedding setup
- even for integer problem

optimal embedded  
Ising problem



DLR-SC Quantum Computing Software  
[gitlab.com/quantum-computing-software](https://gitlab.com/quantum-computing-software)

# First Insights after Annealing Runs



→ Comparison to the current implementation by D-Wave

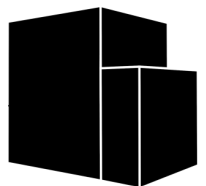
## Random Ising Problems

## Graph Partitioning Problems

- Integer version of OEIP produces significantly better coefficient ratios
  - Best raw solution or solution with best chain break fraction does in general not correspond to best de-embedded solution
  - Certain embeddings seem to produce higher chain breaks than others for diverse instances, in particular the clique embedding

- Very slightly reduced chain break fraction
- Significantly worse best objective values

- Significantly improved chain break fractions
- Slightly improved best objective values



# Imprint



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Institute: German Aerospace Center (DLR) – Institute of Software Technology (SC)  
Date: 2025-03-14  
Image sources: All images “DLR (CC BY-NC-ND 3.0)” unless otherwise stated