

# Ensuring relevance of benchmarks for Quantum Computing

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# Agenda

## Our approach to Quantum Computing

- Pasqal & Neutral-atom QPU

## Accounting for different paradigms

- From NA-QPUs to graphs
- From Q. Simulation to Analog QC
- Hybrid

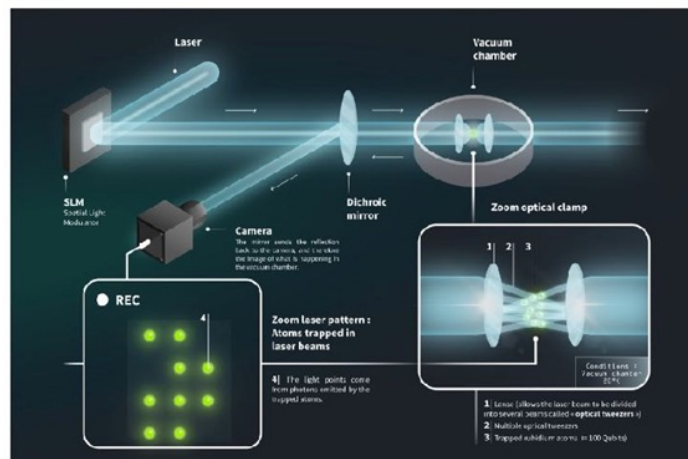
## Building relevant benchmarks

- Choosing the right metrics
- Example 1 : Column Generation (Optimization)
- Example 2 : Quantum Kernel (Machine Learning)

## Conclusion

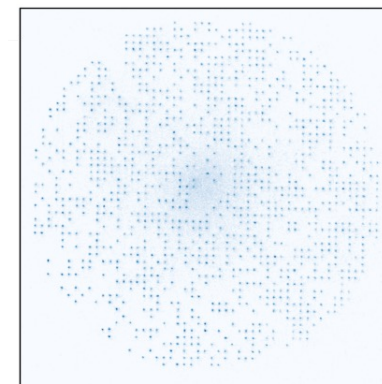
# Our approach to Quantum Computing

- Founded in 2019
- 200+ scientists & engineers
- Building neutral-atom QPUs



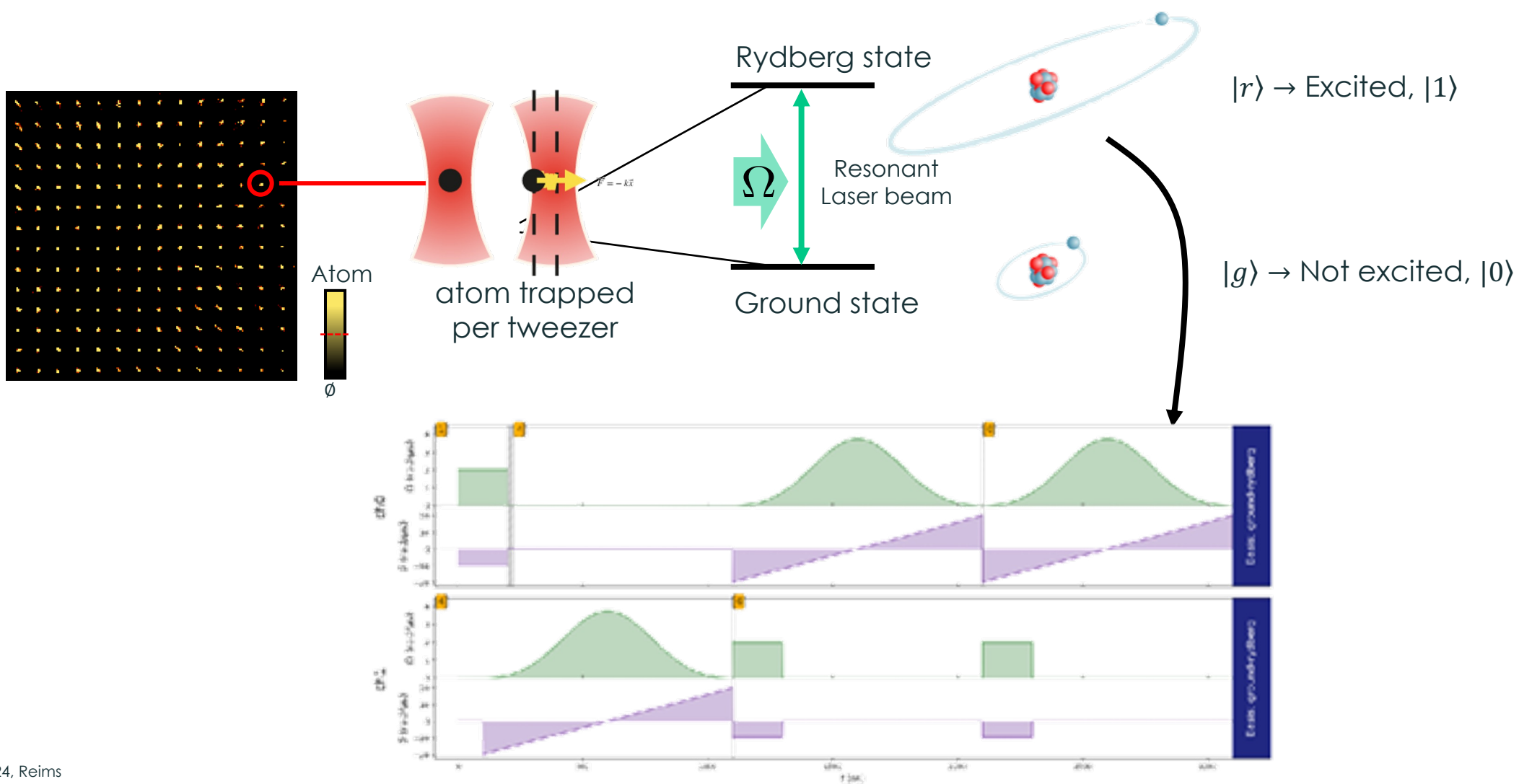
- Full stack : from HW to applications

- 1 industrialized QPU on the cloud
- 2 to be installed in HPC centers
- More on the way
- Electrical consumption of 4 washing machines



- 2 devices dedicated for R&D
  - Larger qubit number
  - FTQC program

# Neutral atom QPU in a nutshell

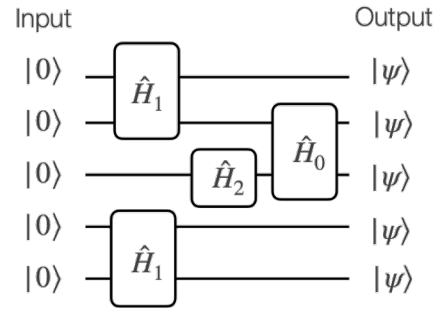


# Modes of operation

## Gate-based

Programming a quantum circuit with quantum gates

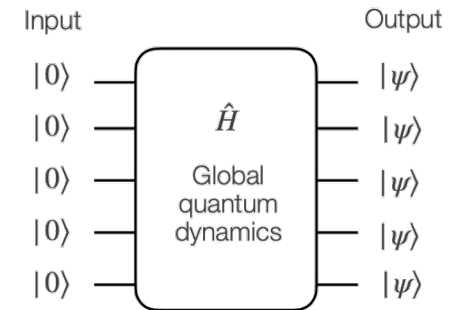
Elementary operations are discrete digital quantum gates, that can act either on individual qubits, or on several qubits at the same time.



## Analog

Programming a Hamiltonian sequence

The Hamiltonian faithfully describes the dynamics of a physical quantum system or a reformulation of an operational case. Parameters can be tuned continuously.



gg-qubit

$|g\rangle \equiv |0\rangle \quad |g'\rangle \equiv |1\rangle$

- + Extremely long lifetimes (msec – sec)
- Weakly interacting: requires intermediate  $|r\rangle$  state

➔ 1- and 2-qubit gates

gr-qubit

$|g\rangle \equiv |0\rangle \quad |r\rangle \equiv |1\rangle$

- + Simple, few laser fields
- + Easier to prepare
- + Easier to measure
- Limited by lifetime of Rydberg state ( $\sim 100\mu\text{s}$ )

➔ Ising model

Ishnower (2010), Müller (2014), Maller (2015), Kaufman (2015), ...

Bernien (2017), Lienard (2018), Guardado-Sanchez (2018), ...

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# Accounting for different paradigms

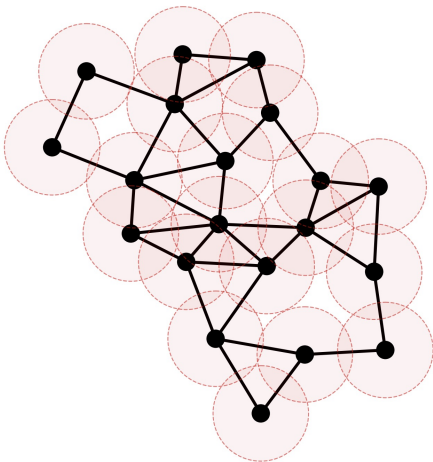
From gates to analog and hybrid

# From Neutral atom QPU to graphs

## Hamiltonian

$$\hat{\mathcal{H}}(t) = \hbar \sum_{i=1}^N \left( \frac{\Omega(t)}{2} \hat{\sigma}_i^x - \delta(t) \hat{n}_i \right) + \sum_{i<j} \frac{C_6}{|\mathbf{r}_i - \mathbf{r}_j|^6} \hat{n}_i \hat{n}_j$$

Unit disk graph

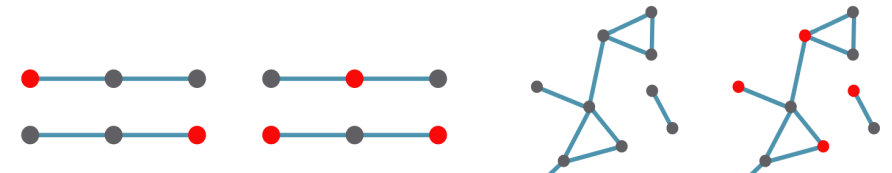


## Graphs

### Maximum Independent Set(s)

Let  $\mathcal{G} = (V, E)$  be a graph.

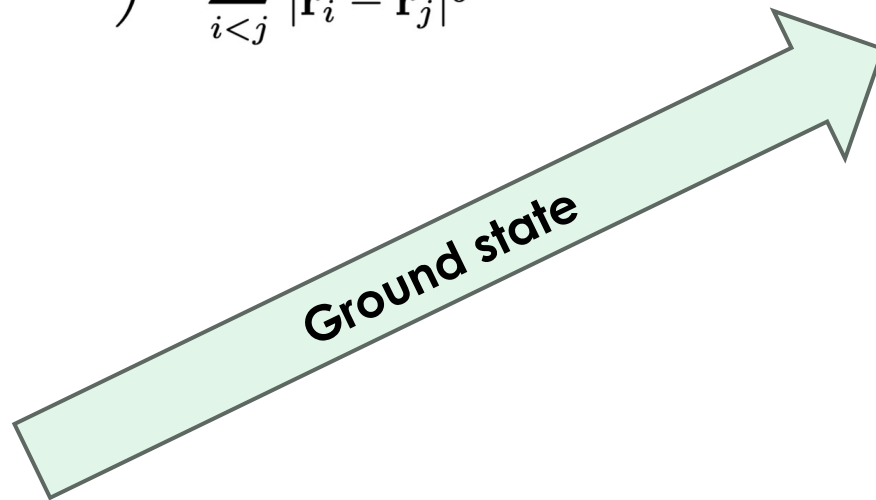
$S \subset V$  is an independent set if  $\forall v, w \in S, (v, w) \notin E$ .



$$H_{\text{MWIS}} = - \sum_i \delta_i n_i + \sum_{i<j} V_{ij} n_i n_j$$

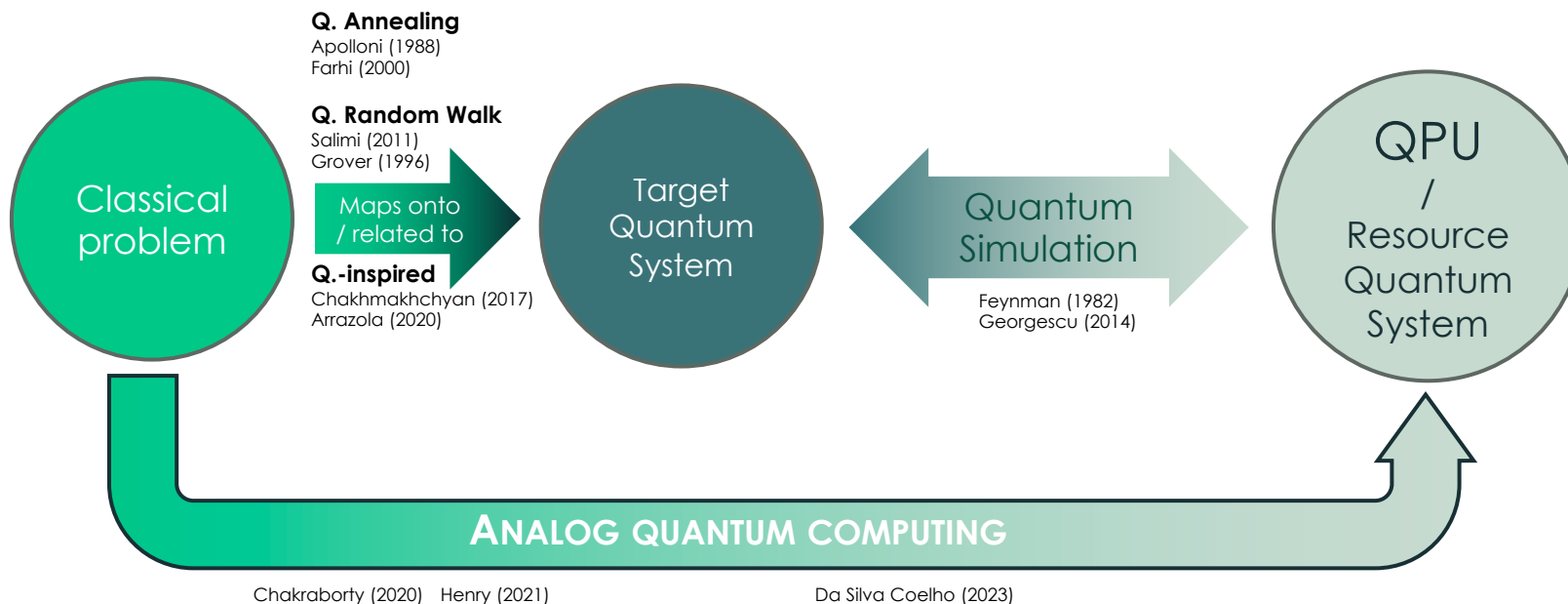
**Optimization problem  
with a wide range of applications**

**Graph-structured problems  
are natively addressable  
with neutral atoms!**





# From Quantum Simulation to analog QC

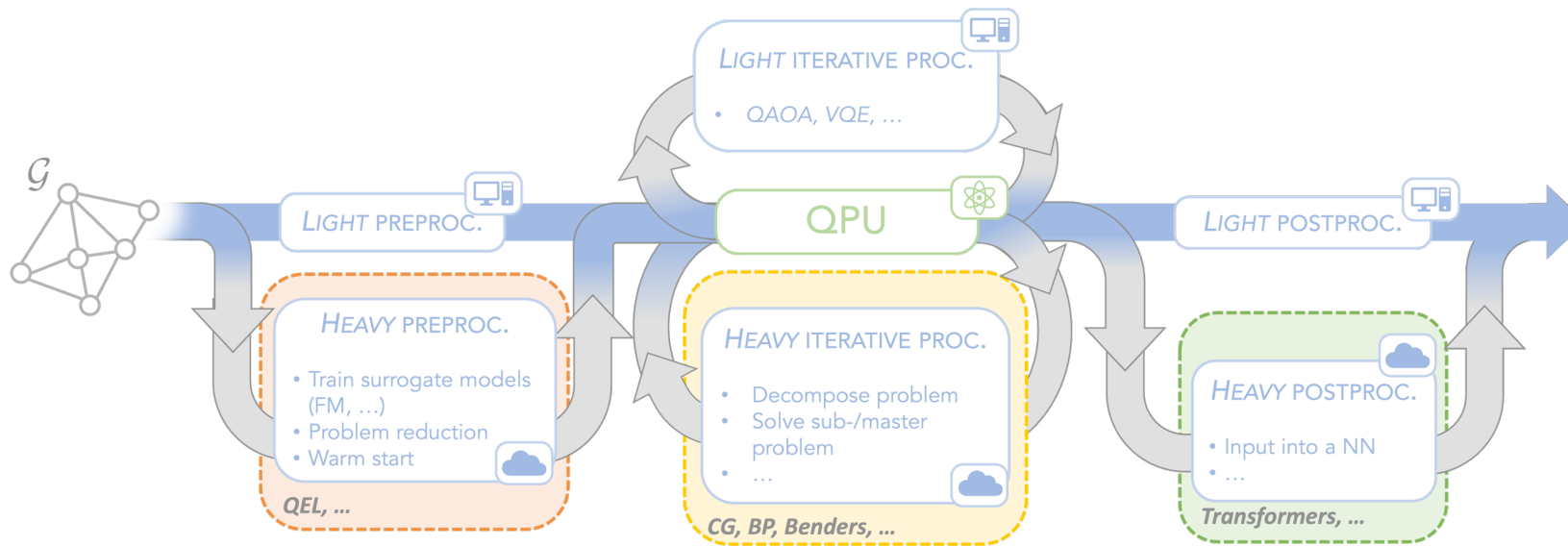


## Analog vs gate-based

- Not a short-term « esatz », but a different paradigm
  - *Some similar algorithms*
  - *New ones*
- Interactions are always on
  - *Quick development of correlation/entanglement*
  - *No need of Trotterization*
  - *Undesired cross-talk*
- Programed through continuous control
  - *Lot of freedom*
  - *No clear framework yet*

## What do we mean by « hybrid »?

Heavily rely on classical processing (Might derive from a fully classical framework, replacing specific éléments)



- *Examples*

- **Optimization** (Column Generation, Branch & Bond, ...)

- **Graph Machine Learning** (Transformers, ...)

## How do you compare ?

- Classical and quantum costs can be similar :
  - *How do you take into account the cost of the classical part?*
  - *What do you compare the quantum elements to?*
- What metric is the most relevant?
  - *→ Application-driven metrics*
- *(existing classical ones, Q-score, ...)*

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# Building relevant benchmarks

To better fit users needs and help development

# Choosing the right metrics

## How one compares Classical and Quantum

- Determining the scaling of an algorithm
  - Classical complexity theory
    - *Nb. of operations, memory usage*
  - Quantum complexity theory
    - *Nb. of gates, circuit depths, Nb. of qubits*
- Advantages of this approach
  - Provides nice theoretical arguments
  - Compatible with classical complexity theory
- Issues with this
  - No clear framework for Q. sim. and analog QC
  - Often restricted to “worst-case scenario” scaling, or to certain types of instances

## What one wants to measure *in practice*

- Most Q. algorithms aim at finding a *bitstring* :  
$$\mathbf{b} = b_0, b_1, \dots, b_{N-1} \in \{0,1\}^N$$
sampled with probability  $p_{\mathbf{b}}$  from the final state
- In practice, one repeats  $n$  times the same preparation and measurement

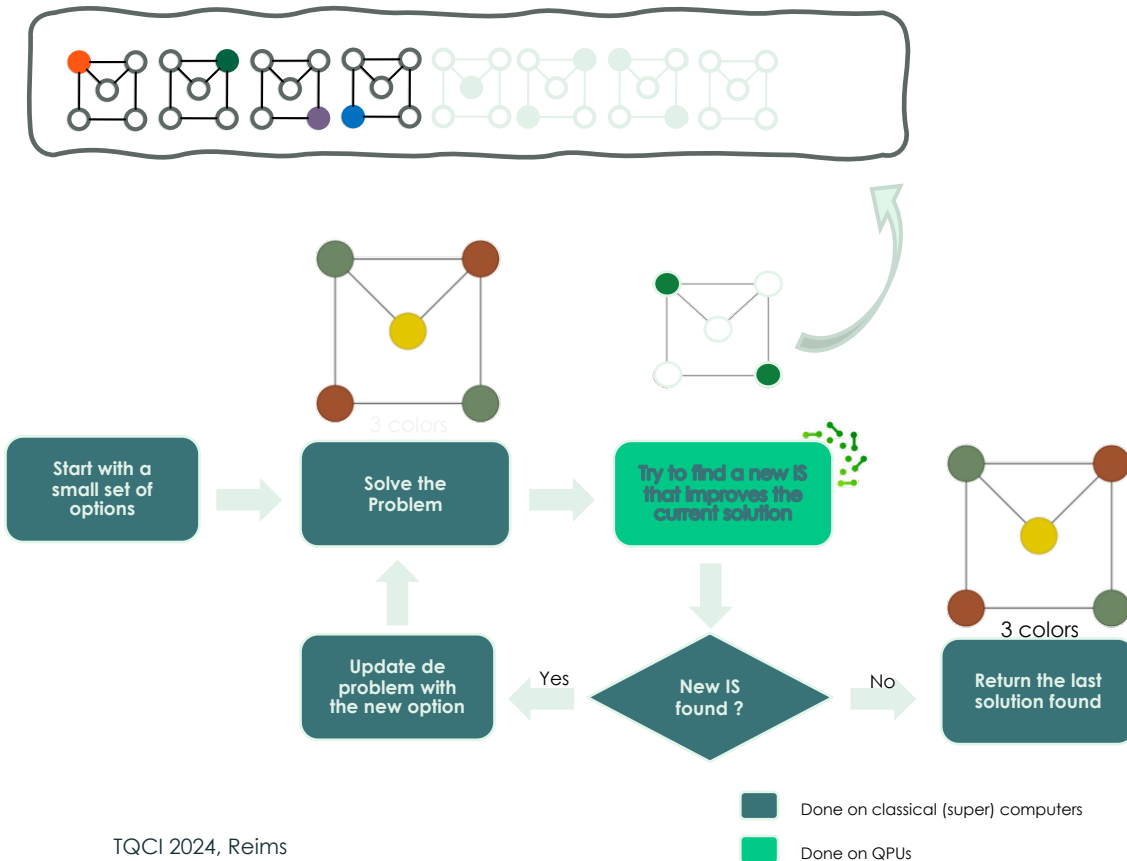
$$\{\mathbf{b}_1, \dots, \mathbf{b}_n\}$$

- For applications, users are rather interested in
  - **Time to solution (TTS)**  $TTS \sim p_{\mathbf{b}}^{-1}$
  - **Quality of solutions**  $\min_i f(\mathbf{b}_i)$
  - **Computation cost/Energy consumption**
- Classical solvers rely a lot on heuristics and are tailored to specific problems

→ **Application-specific benchmarks and metrics**

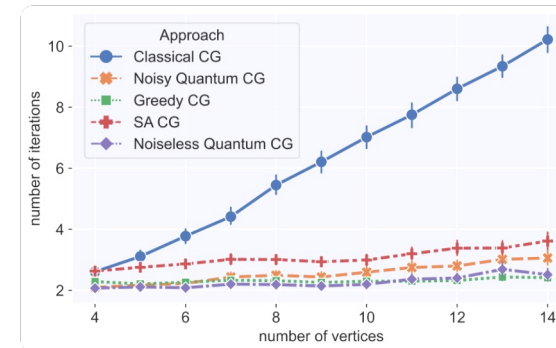
# Example 1 : Column generation (optimization)

## Algorithm

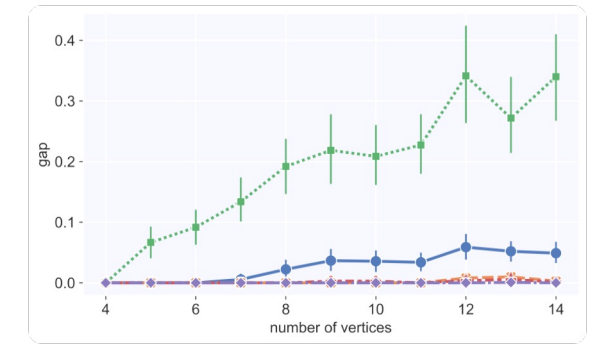


## Results

Number of iterations generating new options



Distance from the global optimal solution

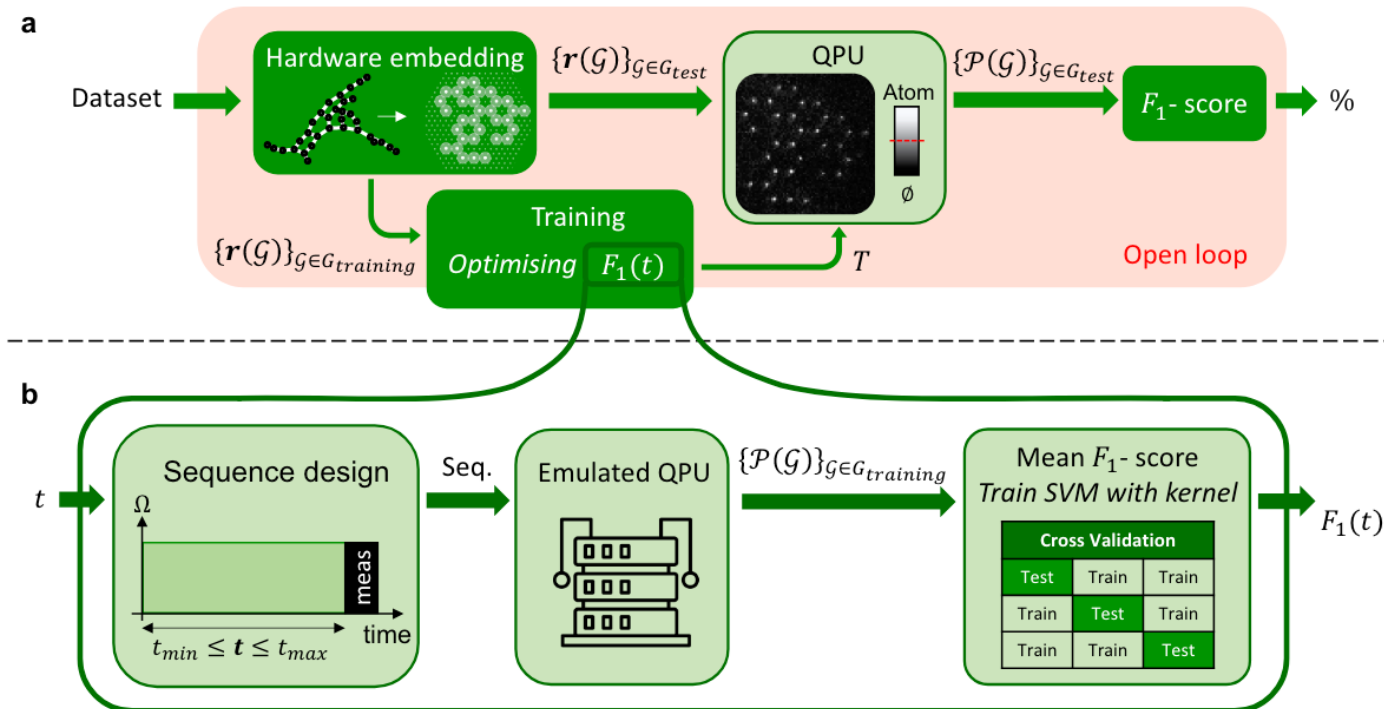


- Our hybrid quantum-classical method has the best overall performance
- It can find better options than state-of-the-art heuristics : up to 80% less colours
- It is faster than fully classical methods : up to 6 times faster than the exact classical framework

# Example 2 : Graph Kernel (Machine Learning)

Experiment on Pasqal hardware:

- Dataset: PTC-FM
- 286 molecules of sizes ranging from 2 to 32 nodes
- Classification task



Kernel	$F_1$ -score (%)
QEK	$60.4 \pm 5.1$
QEK (size-compensated)	$45.1 \pm 3.7$
SVM- $\vartheta$	$58.2 \pm 5.5$
Size	$56.7 \pm 5.6$
Graphlet Sampling	$56.9 \pm 5.0$
Random Walk	$55.1 \pm 6.9$
Shortest Path	$49.8 \pm 6.0$

# Conclusion

## A wide variety of applications

Quantum applications are extremely diverse

- *Different HW technologies*
- *Different QC paradigms :*
  - *Gate-based, analog, annealers...*
- *Different use of QPUs :*
  - *Variational, hybrid...*

This diversity might stay beyond near-term

- *Low-level benchmarks have limited interest*

⇒ *Application benchmarks!*

## Defining Application benchmarks

- Content
  - Relevant and diverse datasets
  - Useful metrics (applicable to various platforms and approaches)
  - Keep up with the classical SotA
  - HW specific as well as HW independant

- How to do this ?

⇒ *Work with classical experts and with end users!*



# Implementing benchmarks



**Technology**  
PASQAL & affiliated ecosystem

## HARDWARE PLATFORM

Max qubits

Addressability

Base repetition rate

FTQC Program

## HARDWARE ACCELERATED LIBRARIES

Quantum Matter & Quantum AI



**Products**

## QUANTUM PROCESSORS

Generation

Total hours of QPU for users

Factories

## COMMUNITY

Platform

Open-source Software Stack

	2022 - 2023	2024 - 2025		2026 - 2027		2028+
Max qubits	200	1,000		10,000		
Addressability	Z add	Z+X add	Addressable 1Q and 2Q gates			
Base repetition rate	1 Hz	3 Hz		10 Hz		100 Hz
FTQC Program		Atom shuttling	Ultra High-Fidelity Gates	Scalable logical qubits architecture		
Quantum Matter & Quantum AI	Algorithm Blueprint	Algorithm Development		Production		
Generation	<b>Orion Alpha</b> ~3M gates	<b>Orion Beta</b> ~5M gates On premise delivery	<b>Orion Gamma</b> ~10M gates On premise delivery	<b>Vela</b> ~40M gates	<b>Pegasus</b> ~200M gates	<b>Centaurus FTQC QPU</b> 128+ Logical qubits 200M+ gates
Total hours of QPU for users	500	5-10,000	20-30,000	60-70,000	200-250,000	500-550,000
Factories	France	Canada	Factory 3			
Platform		Learn	Interact	Collaborate		
Open-source Software Stack	Pulser	Qadence	Solvers & Emulators			

Thank you!