

ACAT2024, Stony Brook, March 11th

Quantum simulation with just-in-time compilation

Andrea Pasquale, on the behalf of the QiboTeam



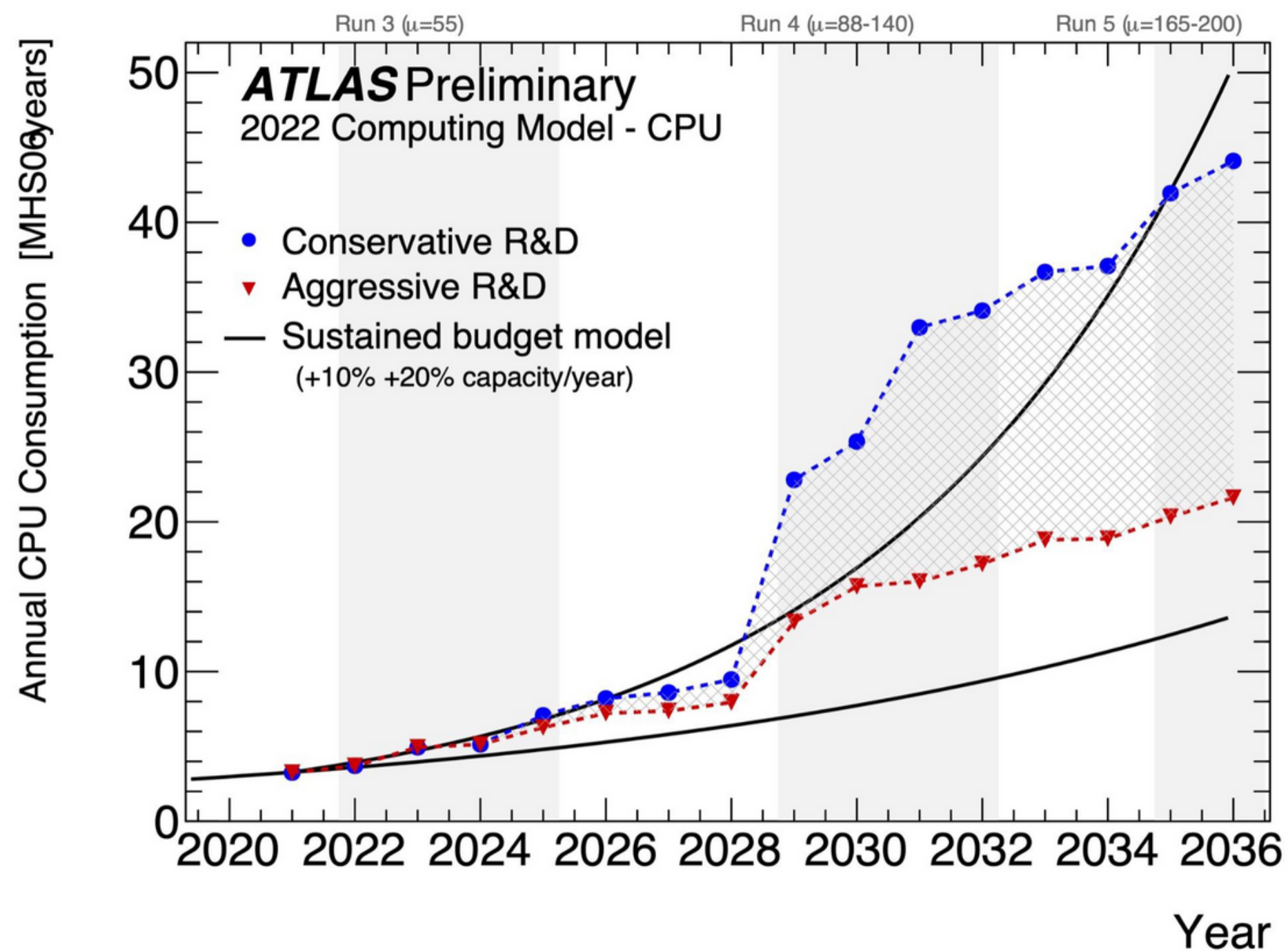
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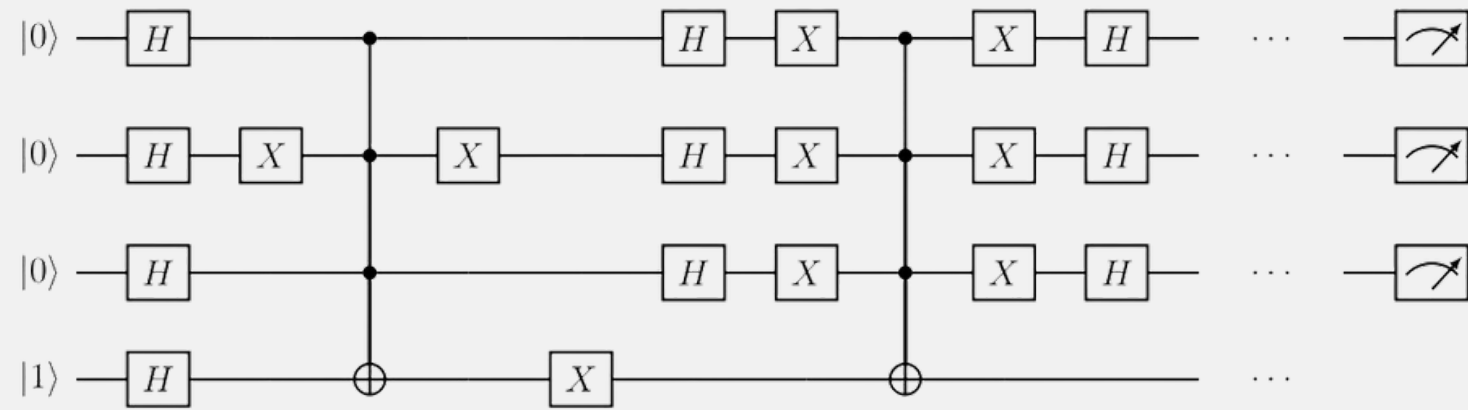
Why people in physics should care about quantum computing?

Someone is already trying...

- A. Perez et al, [arXiv:2011.13934](https://arxiv.org/abs/2011.13934)
- C. Bravo-Pietro et al, [arXiv:2110.06933](https://arxiv.org/abs/2110.06933)
- J. Cruz-Martinez et al, [arXiv:2308.05657](https://arxiv.org/abs/2308.05657)
- F. Rehm et al, [arXiv:2307.05253](https://arxiv.org/abs/2307.05253)
- M. Robbiati et al, [arXiv:2311.05680](https://arxiv.org/abs/2311.05680)



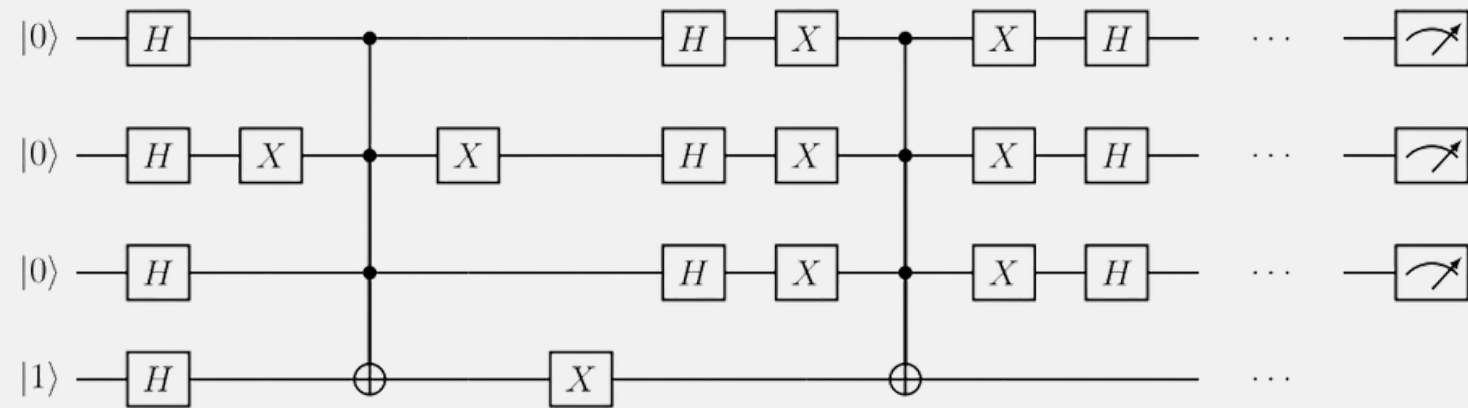
Introduction to Quantum Computing



Example of a Grover algorithm

Why do we care about simulating quantum computers?

Introduction to Quantum Computing

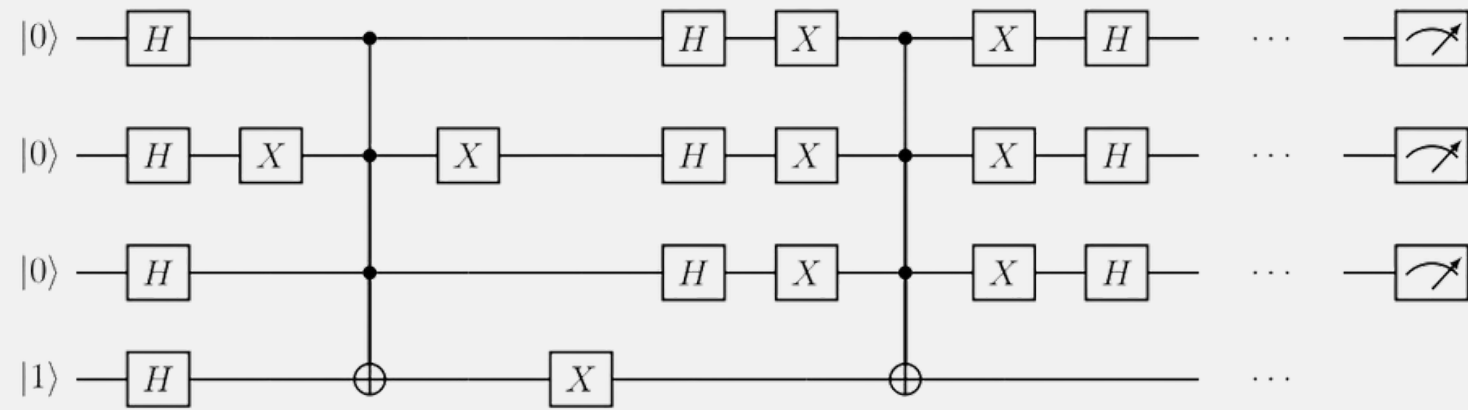


Example of a Grover algorithm

Why do we care about simulating quantum computers?

- Classical simulation is fundamental to understand and design quantum hardware
- To test and verify quantum algorithms before quantum hardware is ready

Introduction to Quantum Computing



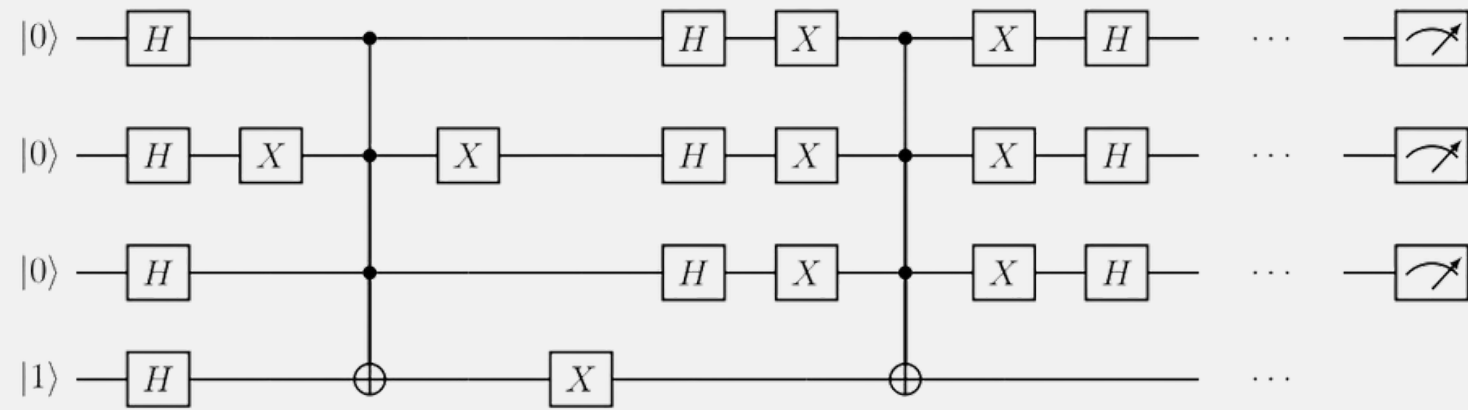
Example of a Grover algorithm

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Simulating quantum computers is hard

Introduction to Quantum Computing



Example of a Grover algorithm

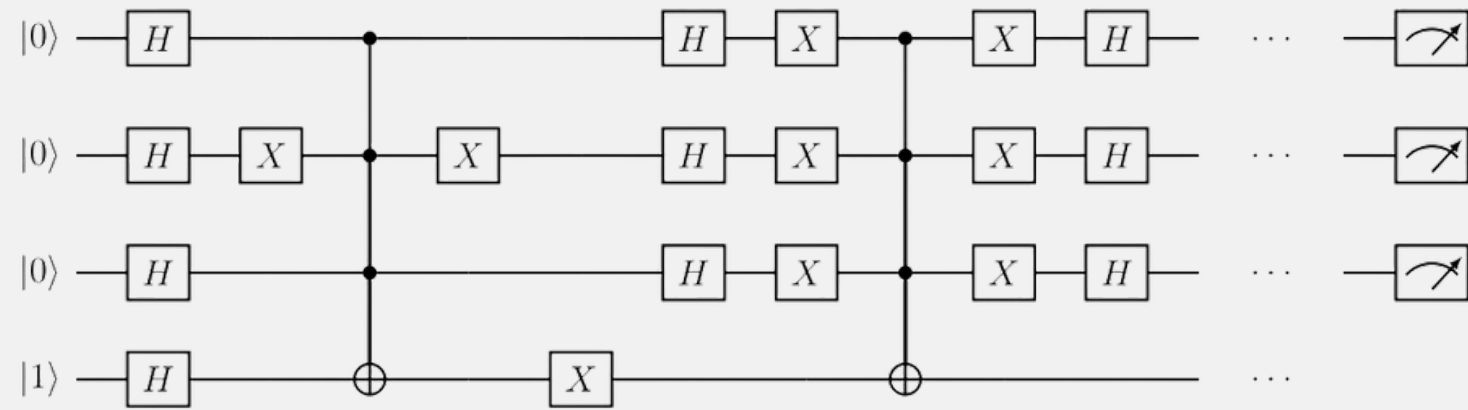
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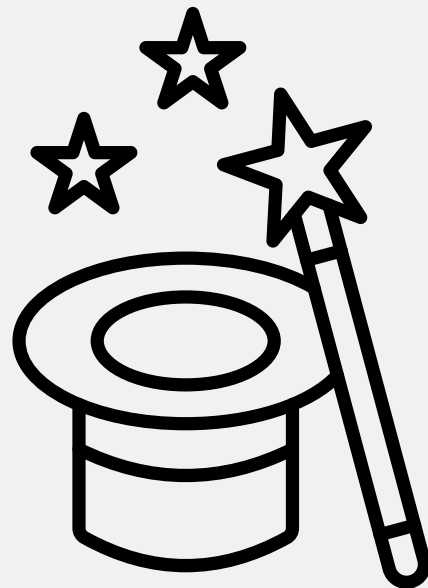
Simulating quantum computers is hard

- Memory management
- Operation needs to be optimized
- Needs to explore different architectures

Introduction to Quantum Computing



Example of a Grover algorithm



Therefore, we need a framework that has some “tricks” to efficiently perform quantum simulation

Why do we care about simulating quantum computers?

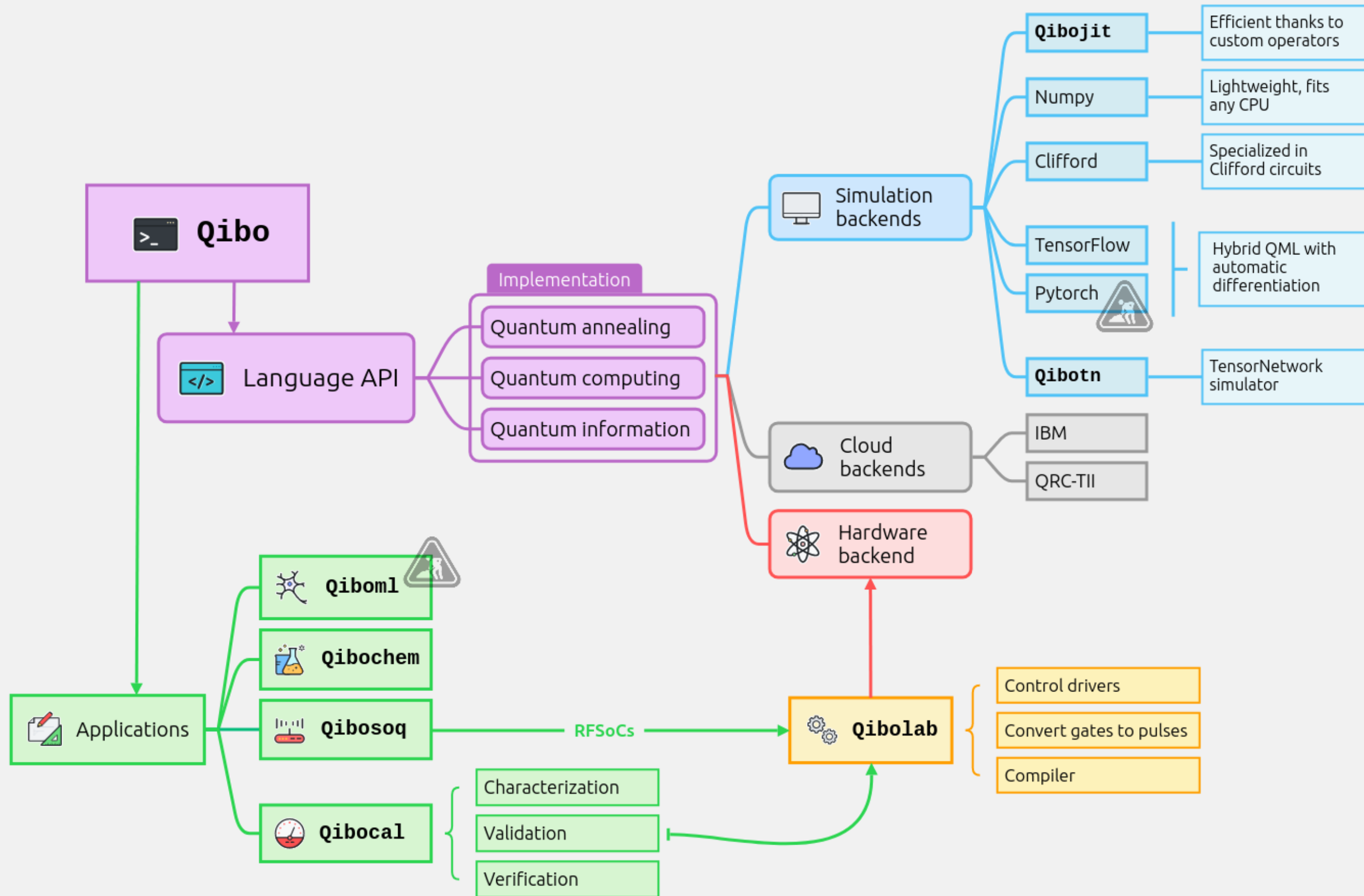
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Qibo

A quantum computing framework for simulation and hardware execution



Qibojit

Full state vector simulation with just-time-compilation

<https://github.com/qiboteam/qibojit>

Full state vector simulation in a nutshell

- N qubit system is represented by 2^N complex numbers
- Deploying a quantum computing algorithms means to apply unitary operators via matrix multiplication
- A good simulator corresponds to a good engine to perform linear algebra operations

Full state vector simulation in a nutshell

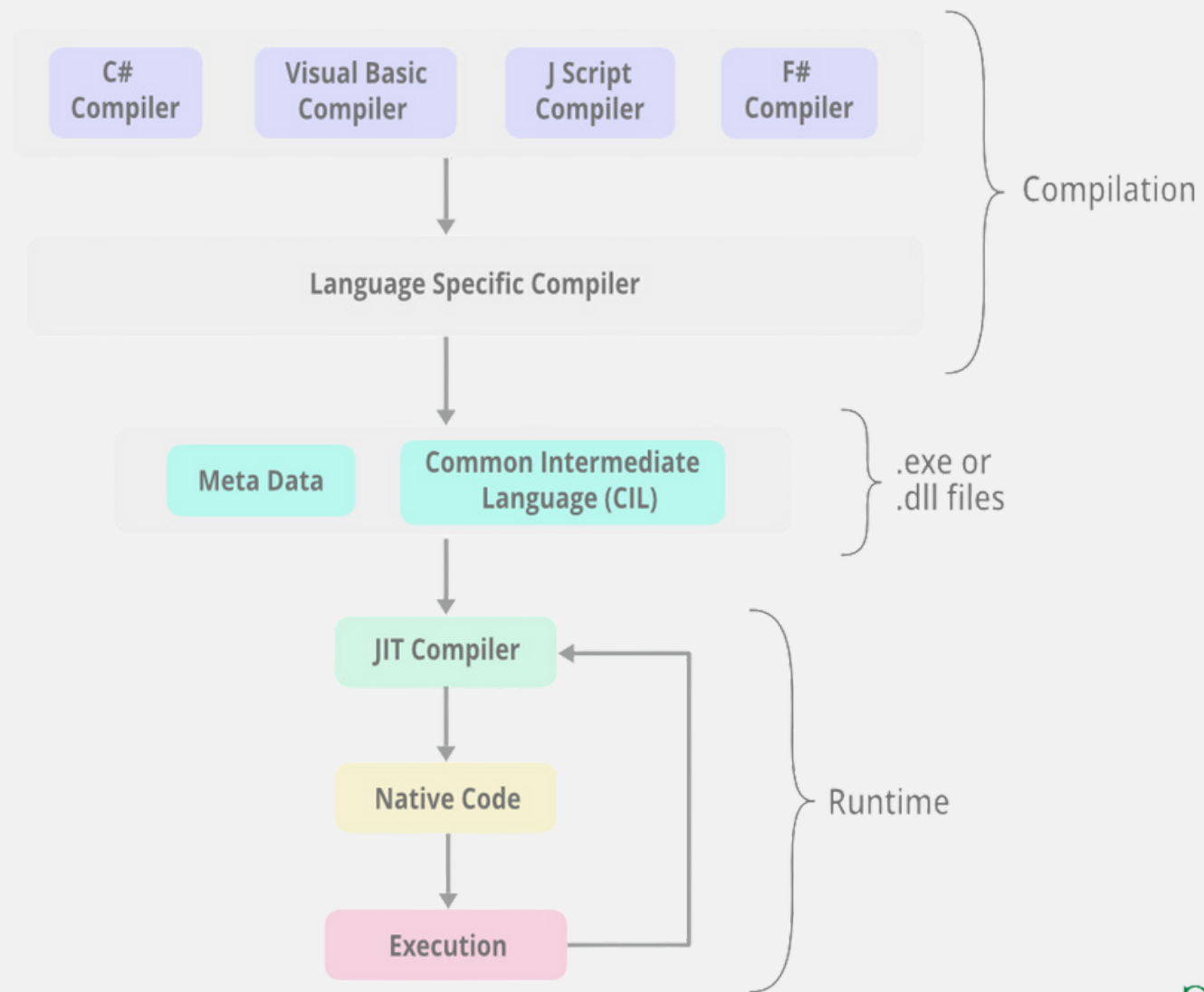
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What can we do to improve performances

- Parallelize using multi threading
- Accelerate through GPUs
- Operation needs to be optimized



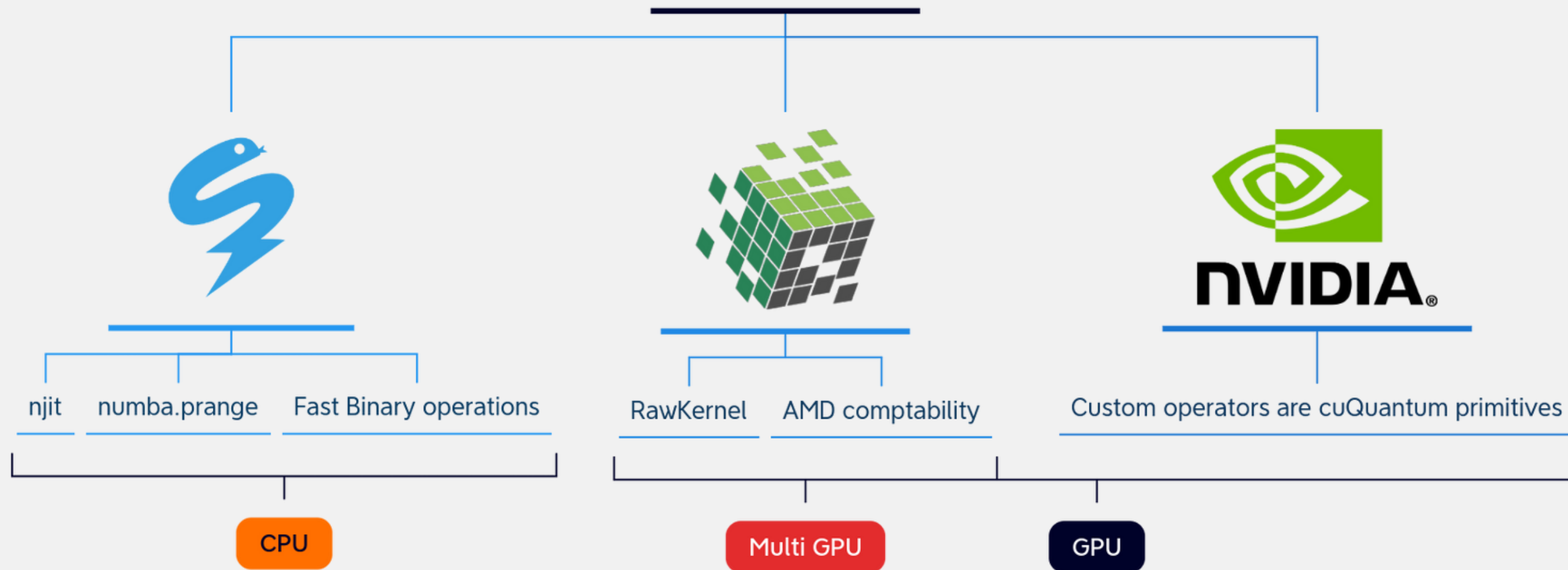
Working of JIT Compiler



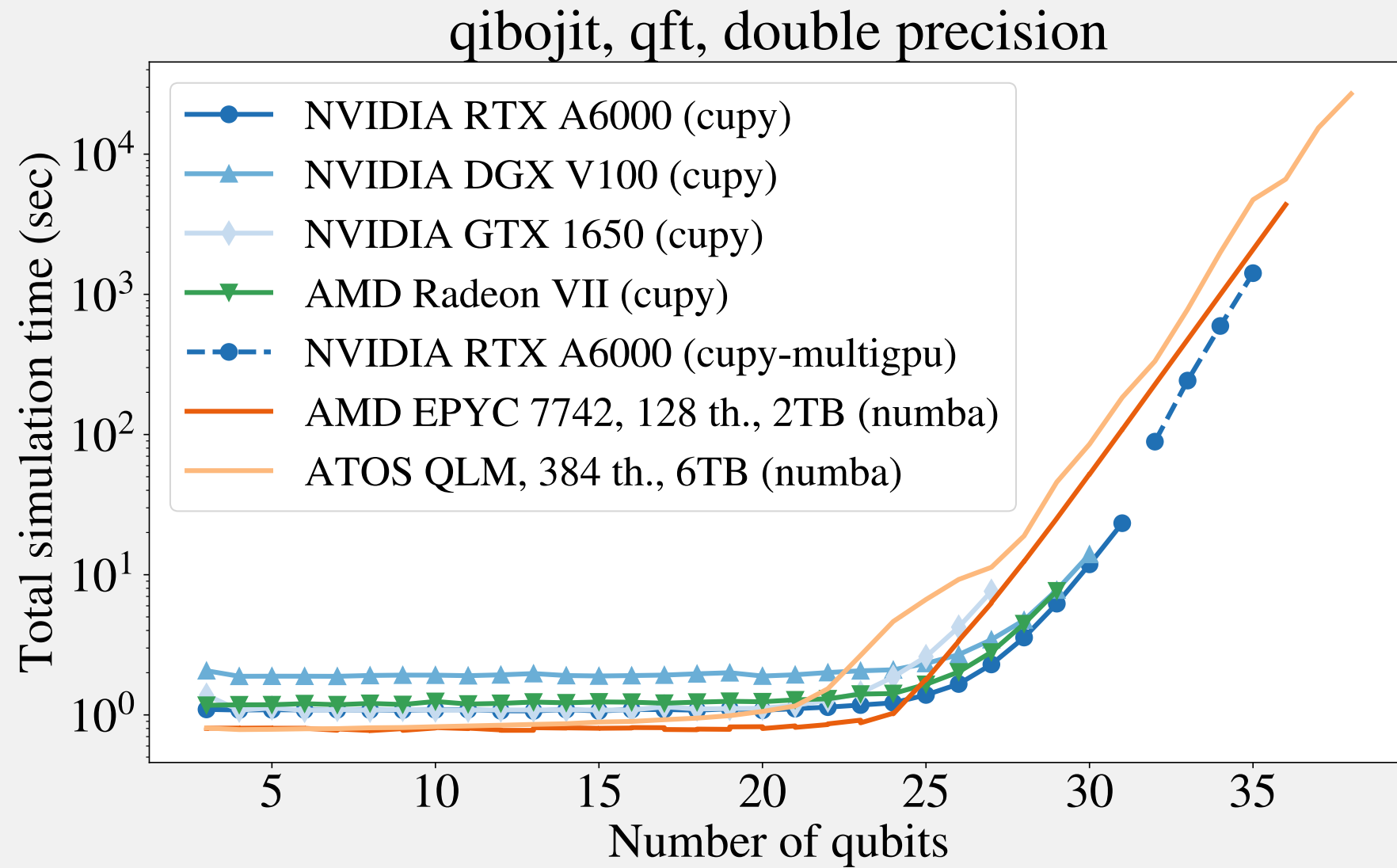
What is just-in-time compilation?

- A method for improving the performance of interpreted languages.
- We can exploit some frameworks in Python to achieve better performances compared to naive approaches

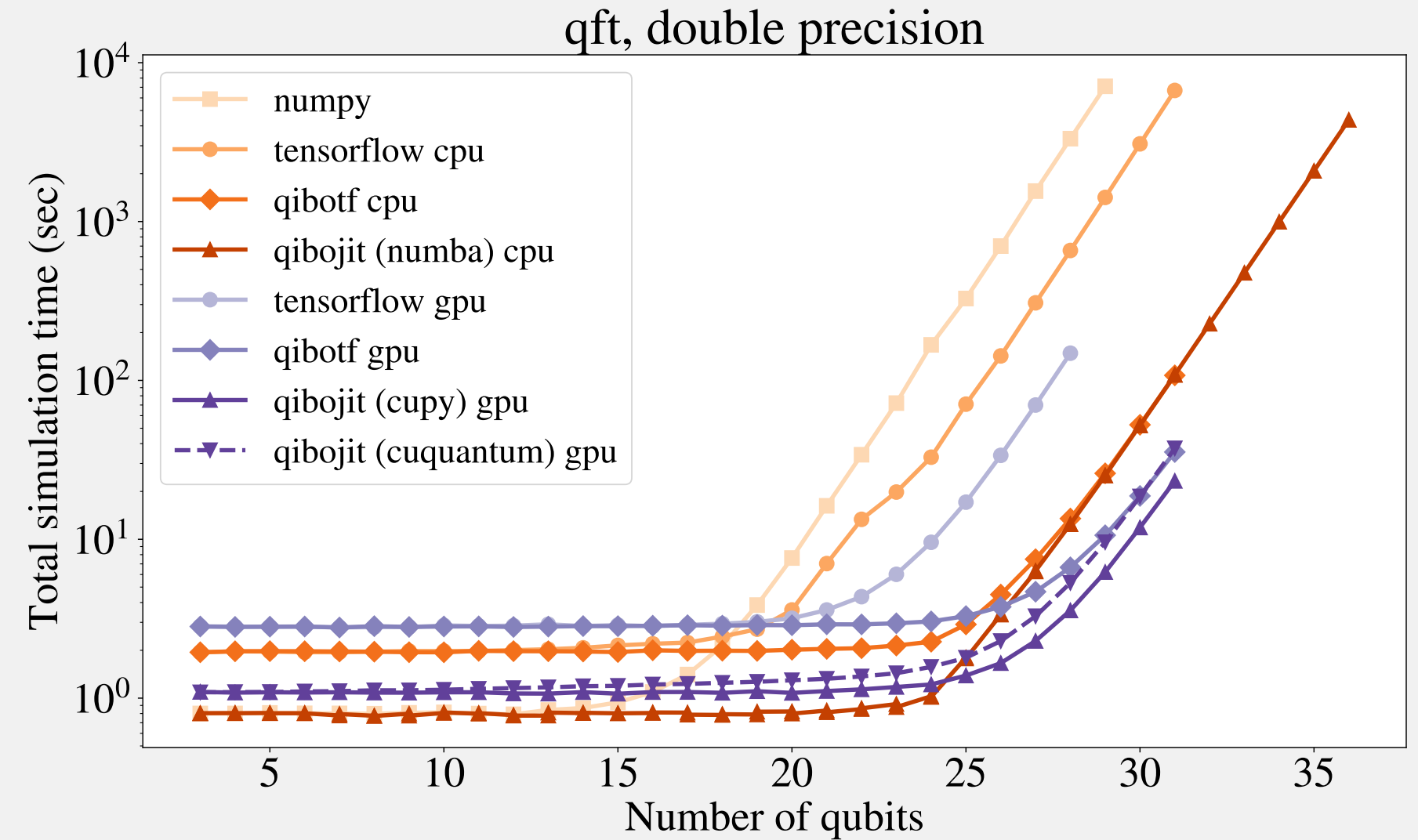
Qibojit



Some benchmarks

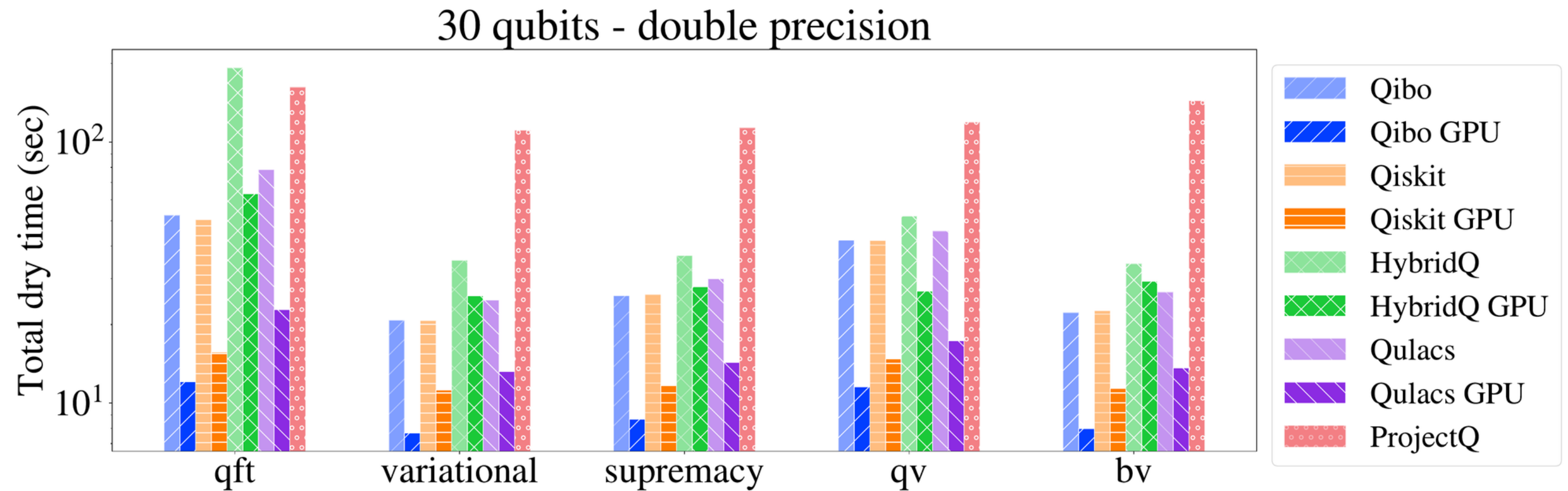


Total simulation time scaling with the number of qubits for simulating the qft circuit on different devices.



Total simulation time scaling with the number of qubits for simulating the qft circuit using different Qibo backends

Comparison with other libraries



Total dry run time for simulating different circuits of 30 qubits, using libraries that support single (top) and double (bottom) precision.

TensorNetworks

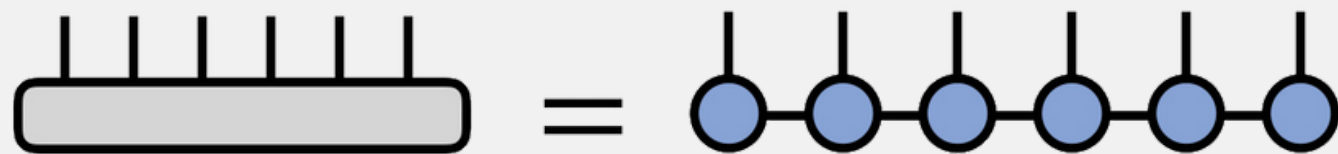
Approximate circuit execution using qibotn

<https://github.com/qiboteam/qibotn>

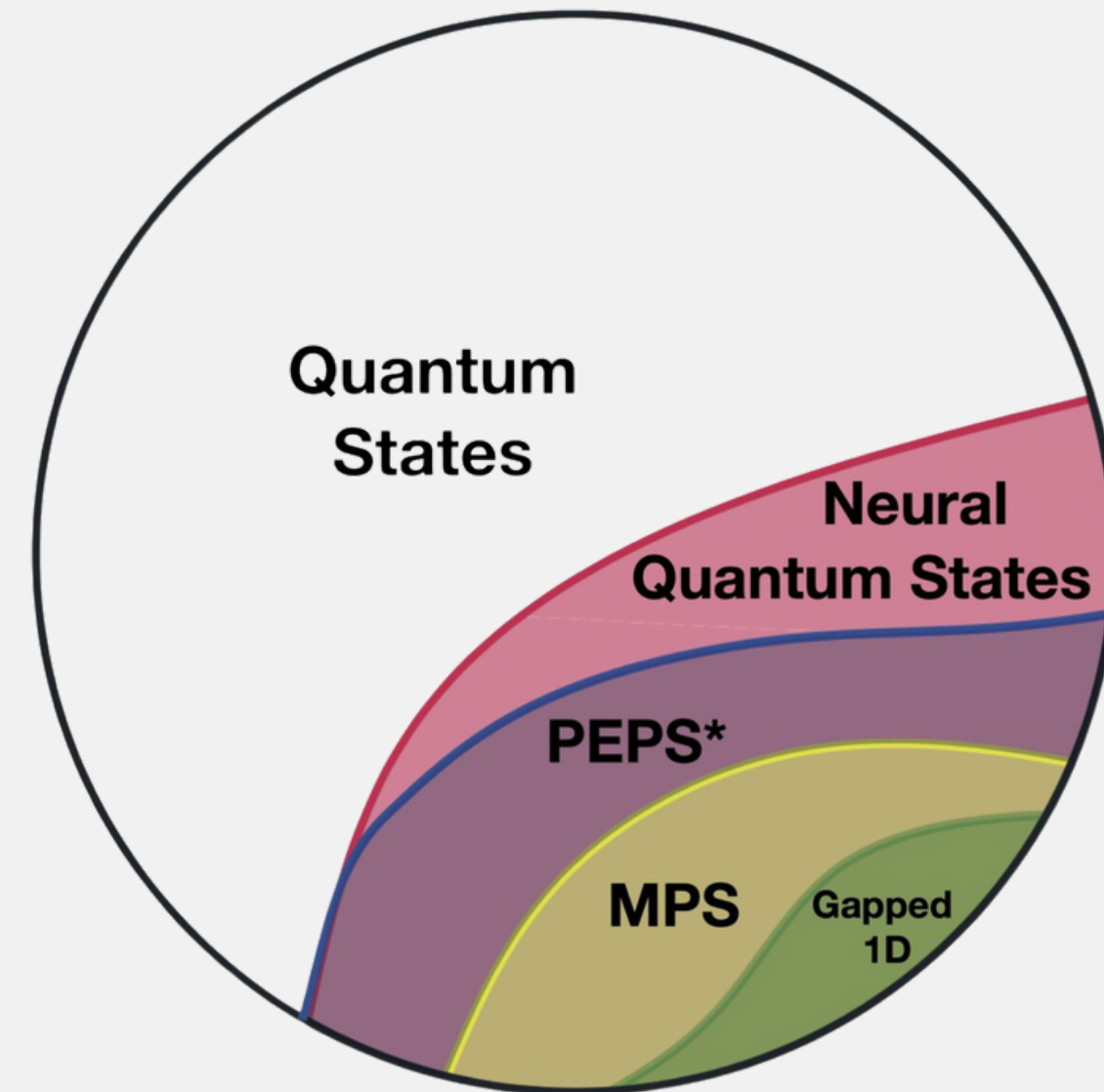
Tensor Networks

Tensor Networks are a powerful method for simulating quantum circuits by representing the state or operator as a network of smaller tensors.

$$|\psi\rangle = \sum_{\sigma_1, \dots, \sigma_L} A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_{L-1}} A^{\sigma_L} |\sigma_1, \dots, \sigma_L\rangle$$



It is an approximate method that enable to simulate circuits with polynomial complexity.



The expressive power of neural networks in quantum physics. Credits to G. Carleo.

[U. Schollwoeck, Rev. Mod. Phys. 77, 259 \(2005\)](#)

[R. Orus, Annals of Physics 349 \(2014\) 117-158](#)

Qibotn

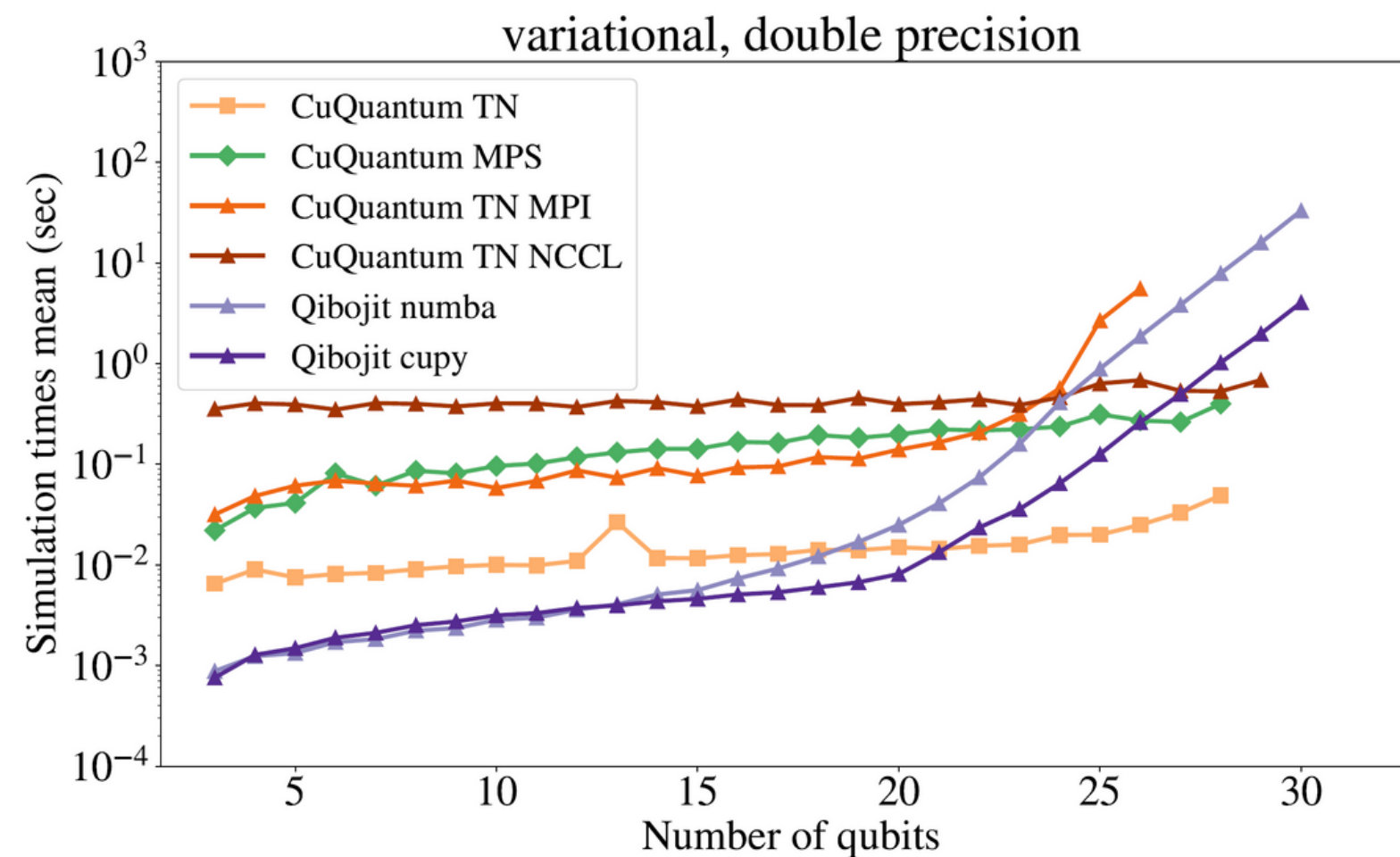
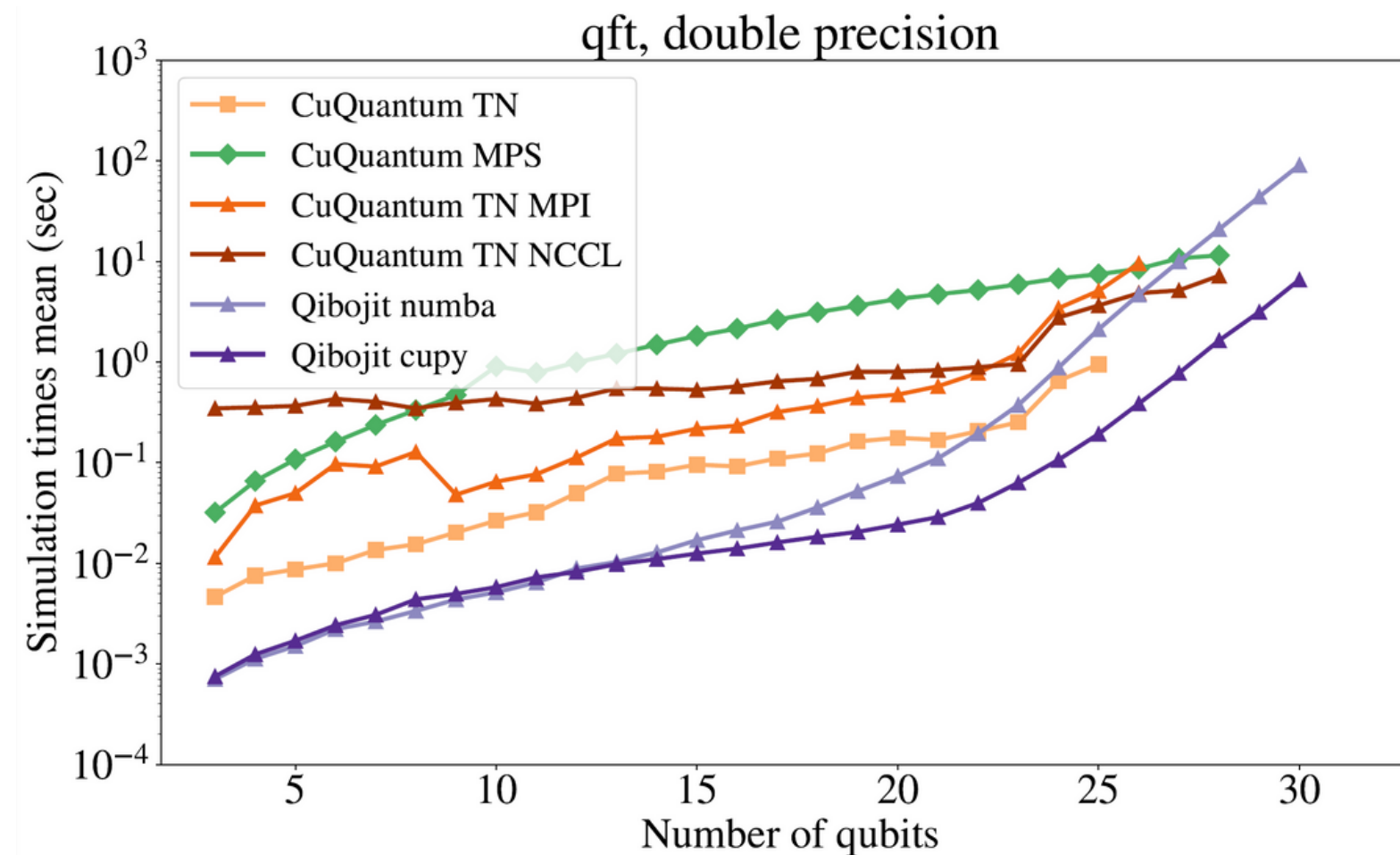


Supported Tensor Networks type:

- TensorNet (TN)
- Matrix Product States (MPS)

Supported HPC configurations:

- single-node CPU
- single-node GPU or GPUs
- multi-node multi-GPU with Message Passing Interface (MPI)
- multi-node multi-GPU with NVIDIA Collective Communications Library (NCCL)

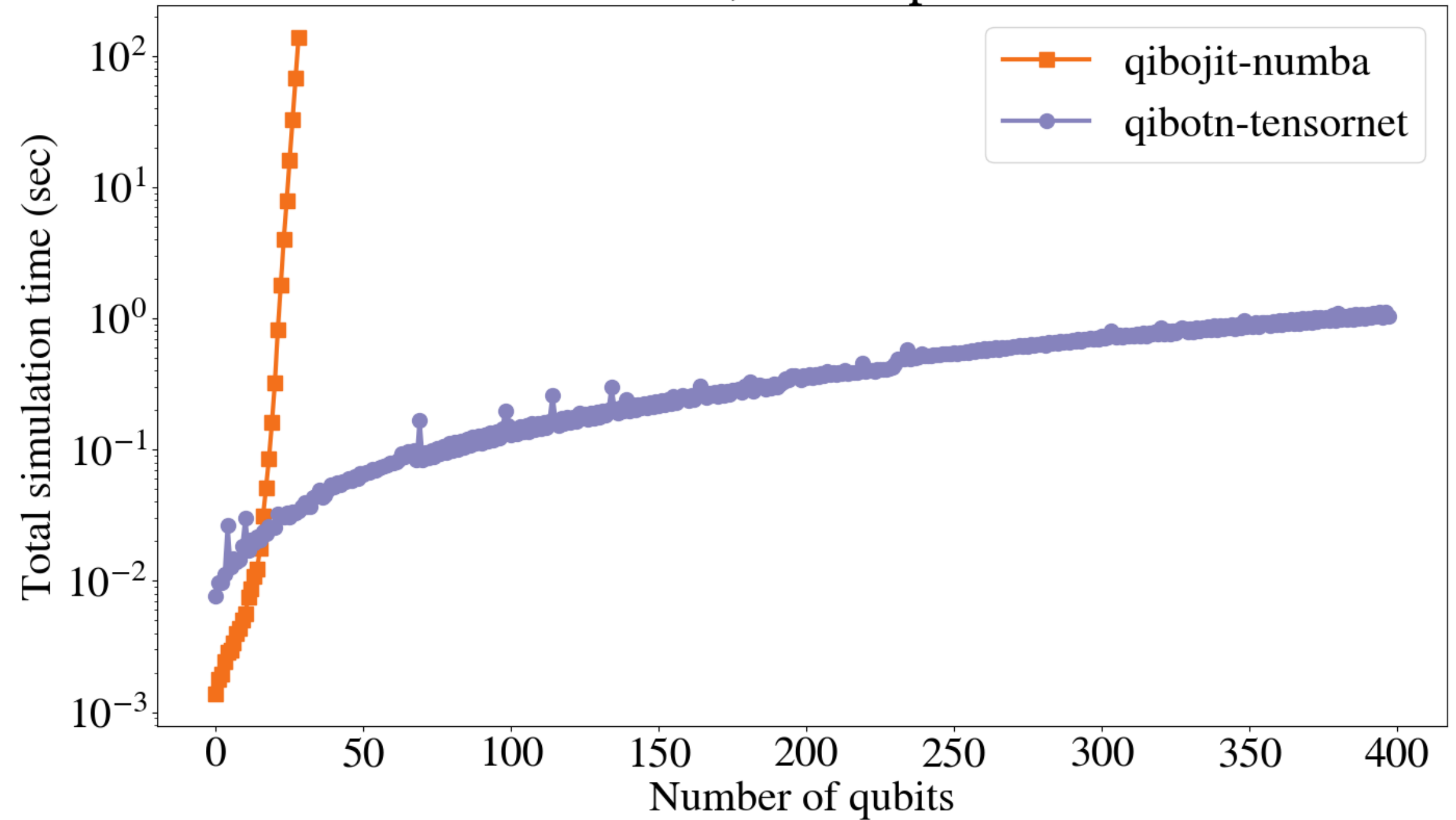


TN can scale up!

Running variational circuits with
400 qubits using A100 GPU



variational, double precision



Clifford simulation

Speeding up execution for Clifford circuits

Clifford Simulation

[Submitted on 25 Jun 2004 (v1), last revised 18 Jun 2008 (this version, v5)]

Improved Simulation of Stabilizer Circuits

Scott Aaronson, Daniel Gottesman

The Gottesman-Knill theorem says that a stabilizer circuit -- that is, a quantum circuit consisting solely of CNOT, Hadamard, and phase gates -- can be simulated efficiently on a classical computer. This paper improves that theorem in several directions. First, by removing the need for Gaussian elimination, we make the simulation algorithm much faster at the cost of a factor-2 increase in the number of bits needed to represent a state. We have implemented the improved algorithm in a freely-available program called CHP (CNOT-Hadamard-Phase), which can handle thousands of qubits easily. Second, we show that the problem of simulating stabilizer circuits is complete for the classical complexity class ParityL, which means that stabilizer circuits are probably not even universal for classical computation. Third, we give efficient algorithms for computing the inner product between two stabilizer states, putting any n-qubit stabilizer circuit into a "canonical form" that requires at most $O(n^2/\log n)$ gates, and other useful tasks. Fourth, we extend our simulation algorithm to circuits acting on mixed states, circuits containing a limited number of non-stabilizer gates, and circuits acting on general tensor-product initial states but containing only a limited number of measurements.

Efficiently simulate gate application and measurements sampling in the stabilizers state representation.

$$\left(\begin{array}{ccc|ccc|c} x_{11} & \cdots & x_{1n} & z_{11} & \cdots & z_{1n} & r_1 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots & \vdots \\ x_{n1} & \cdots & x_{nn} & z_{n1} & \cdots & z_{nn} & r_n \\ \hline x_{(n+1)1} & \cdots & x_{(n+1)n} & z_{(n+1)1} & \cdots & z_{(n+1)n} & r_{n+1} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots & \vdots \\ x_{(2n)1} & \cdots & x_{(2n)n} & z_{(2n)1} & \cdots & z_{(2n)n} & r_{2n} \end{array} \right)$$

The state is represented by a tableau of **binary** variables.

The complexity is:

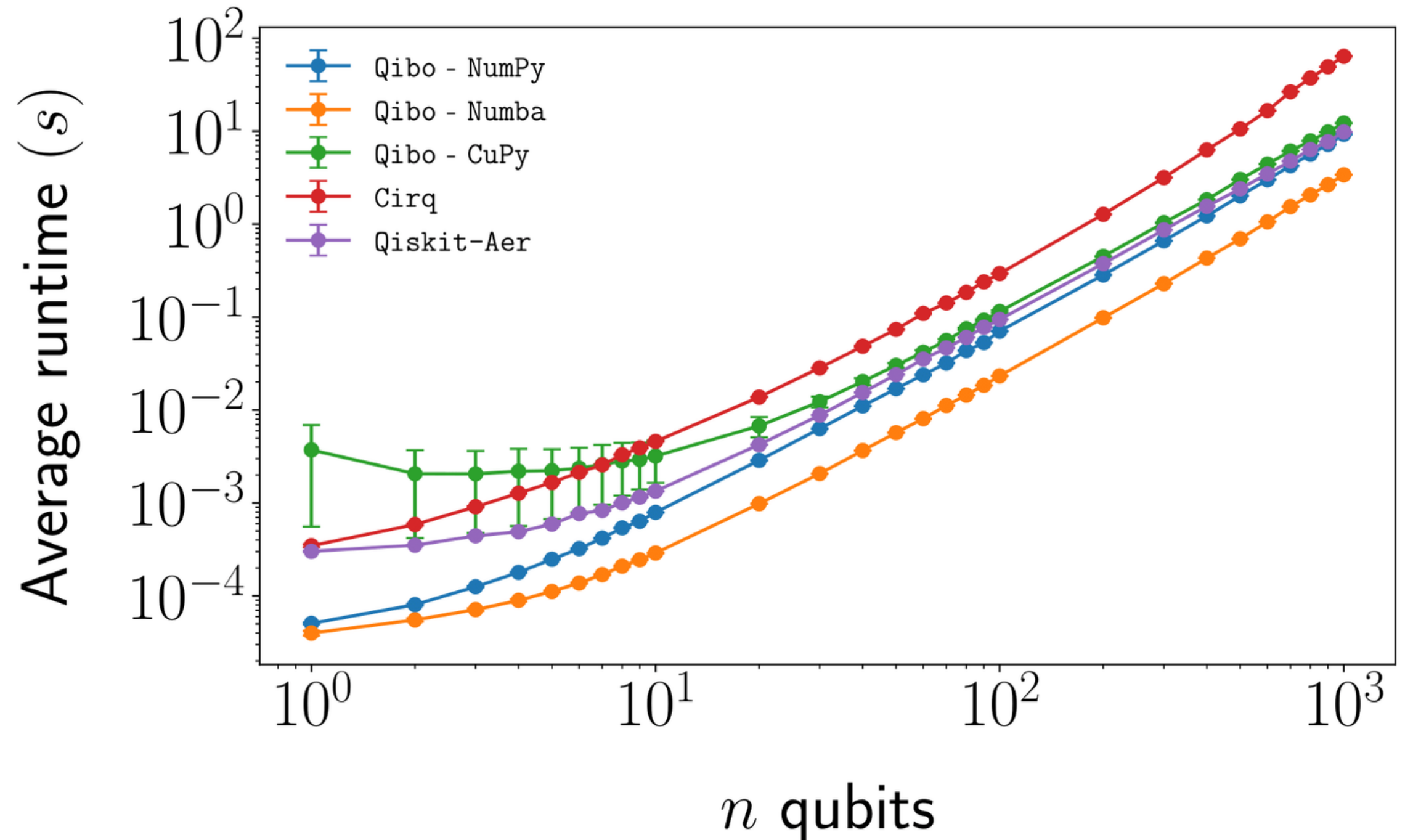
- $O(n)$ for unitary gates
- $O(n^2)$ for measurements

Benchmarks

Simulation of clifford circuits with an increasing number of qubits (no measurements). For each point we take the average over 100 different randomly generated circuits. Each circuit is generated following <https://arxiv.org/abs/2003.0941>, which guarantees an uniform distribution of the generated n -qubits clifford operators, but the depth is not fixed.

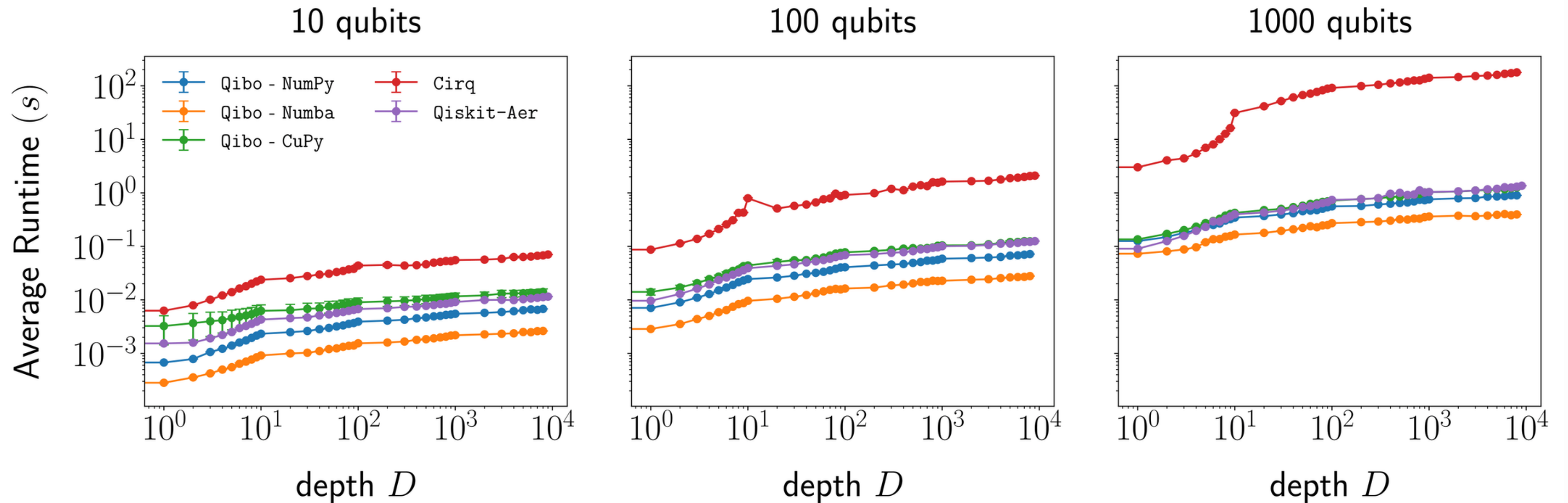
Papaluca et al, In preparation (2024)

Clifford Simulator



Benchmarks

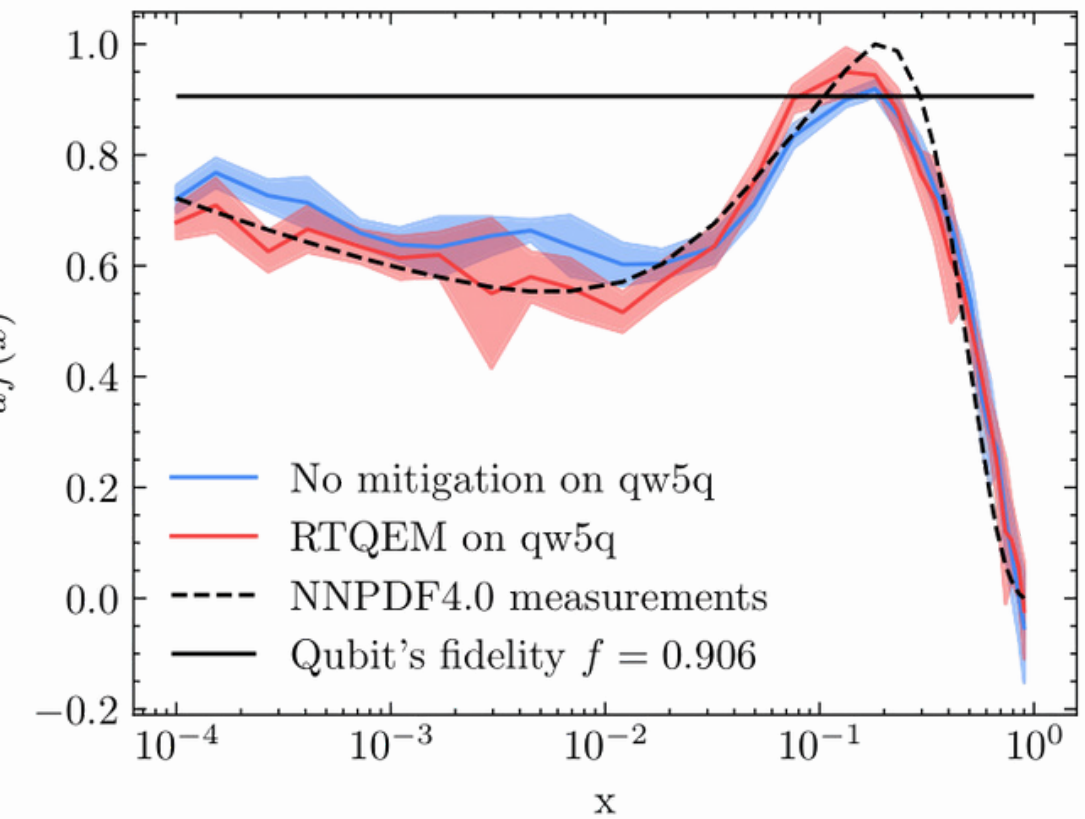
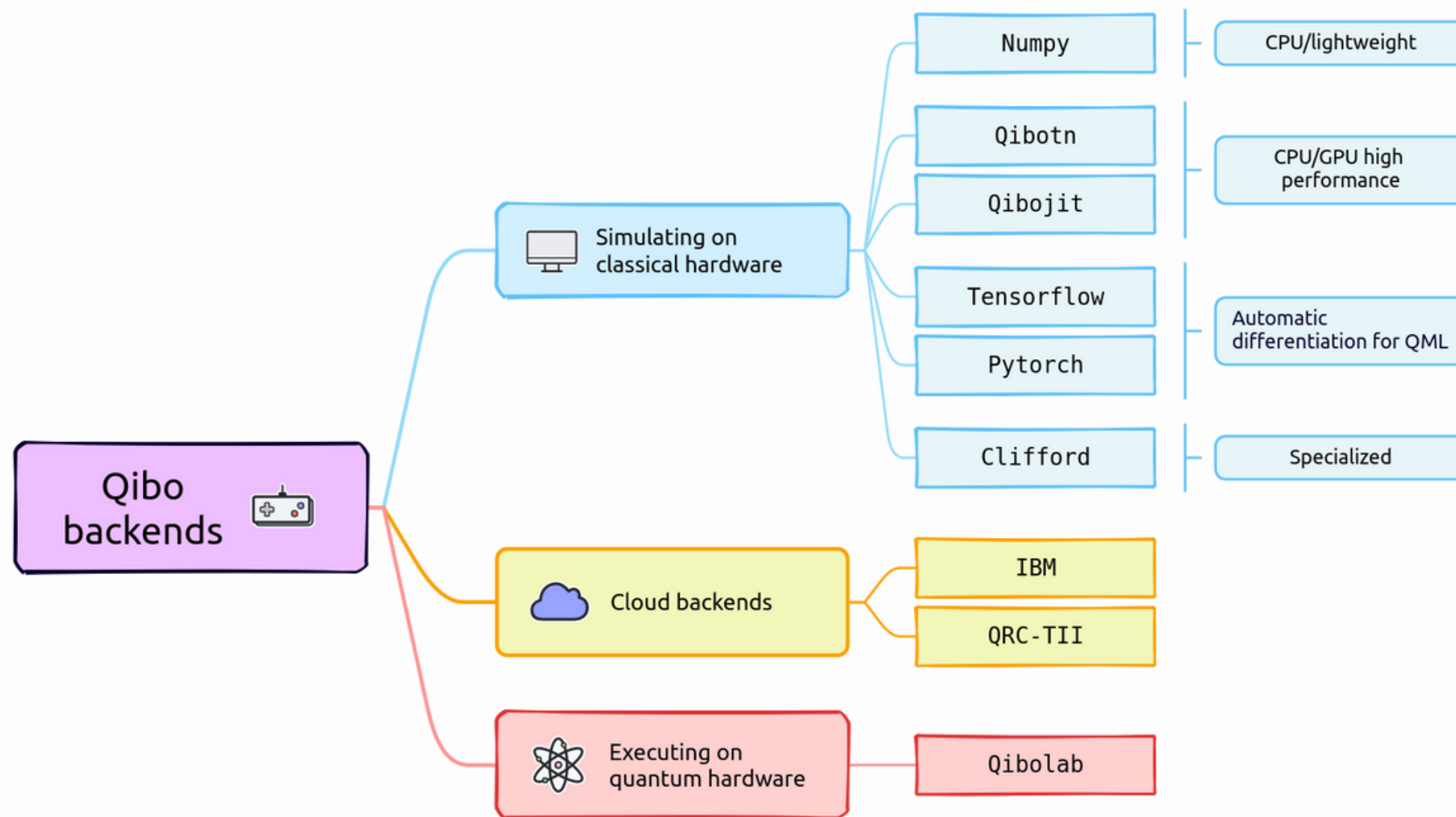
Papaluca et al, In preparation (2024)



Simulation of clifford circuits with an increasing depth for a fixed number of qubits (no measurements). For each point we take the average over 100 different randomly generated circuits. We randomly sample circuit moments composed by a single qubit cliffords layer followed by a two-qubits cliffords one. We sequentially stack up to 10 000 layers constructed this way.

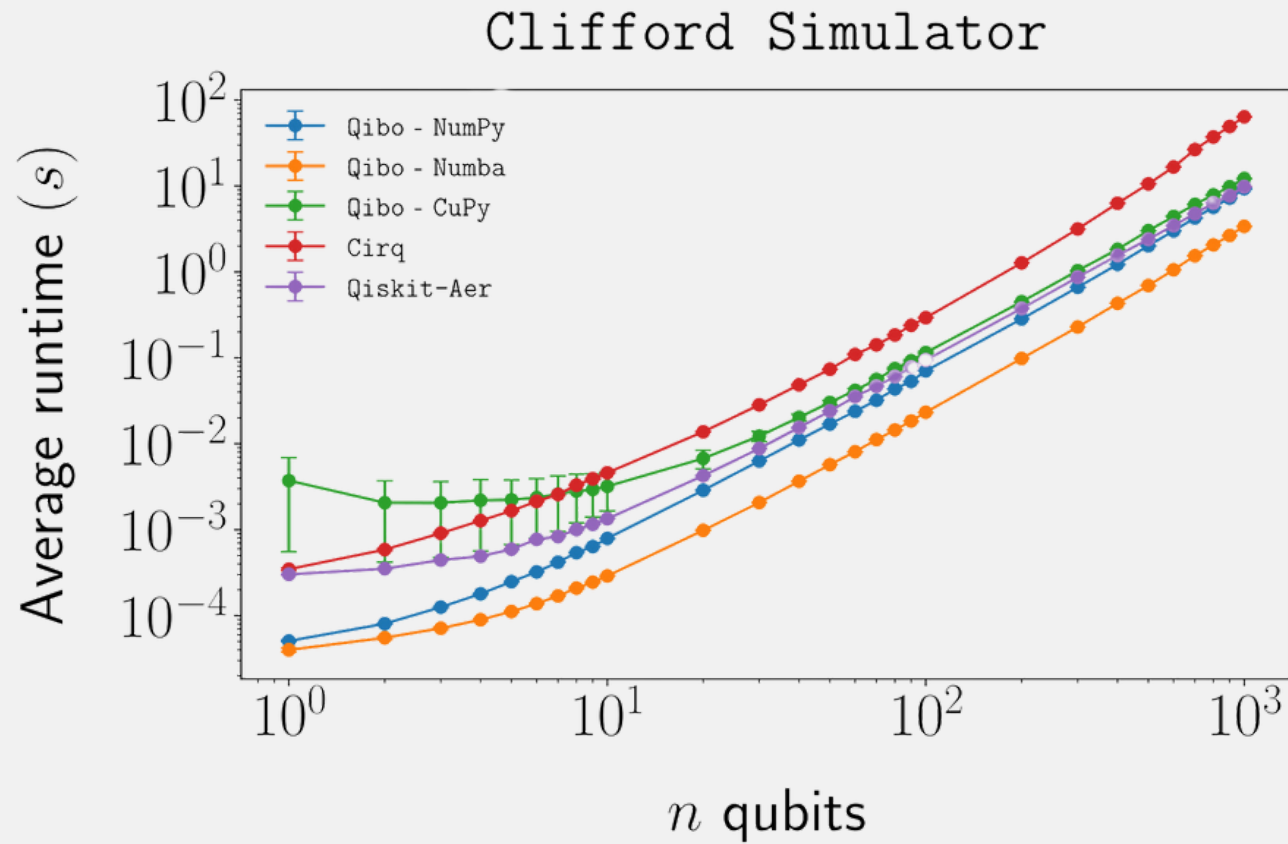
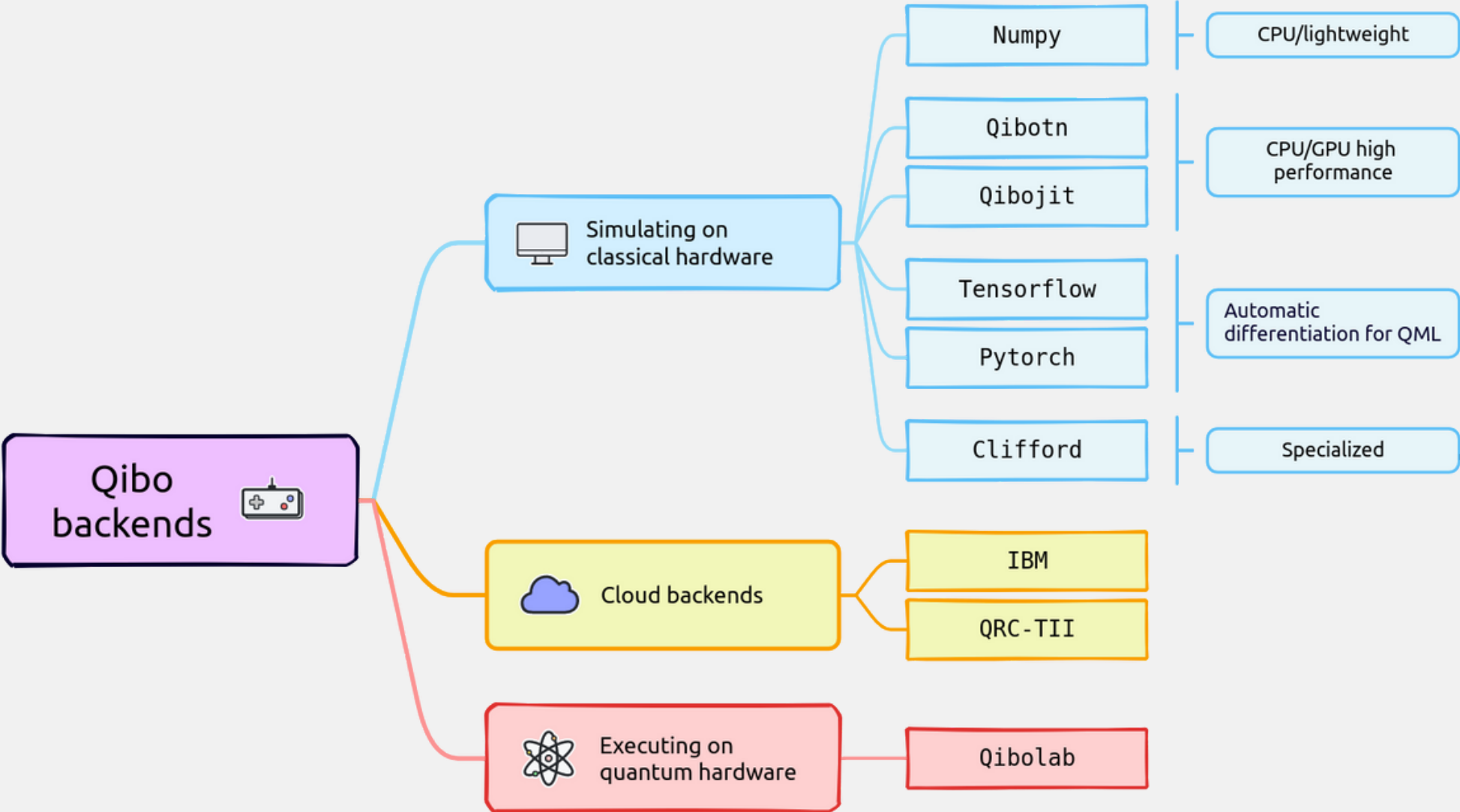
Outlook

Quantum simulation summary



[arXiv:2311.05680](https://arxiv.org/abs/2311.05680)

Quantum simulation summary



We have presented a **fully open source** quantum computing framework.

Qibo is compatible with state-of-the-art quantum simulation and offers several engines:

- **qibojit**: full-state vector simulation
- **qibotn**: tensor-network simulation
- **clifford**: specialized execution

But Qibo is **much more** than a quantum simulators:


- hardware execution and calibration -> Edoardo's talk
- algorithms and error mitigation -> Matteo's talk

New in v0.2.5 New simulation and cloud hardware backends!



Qibo: an open-source middleware for quantum computing

An end-to-end open source platform for quantum simulation, self-hosted quantum hardware control, calibration and characterization.

`$ pip install qibo` 

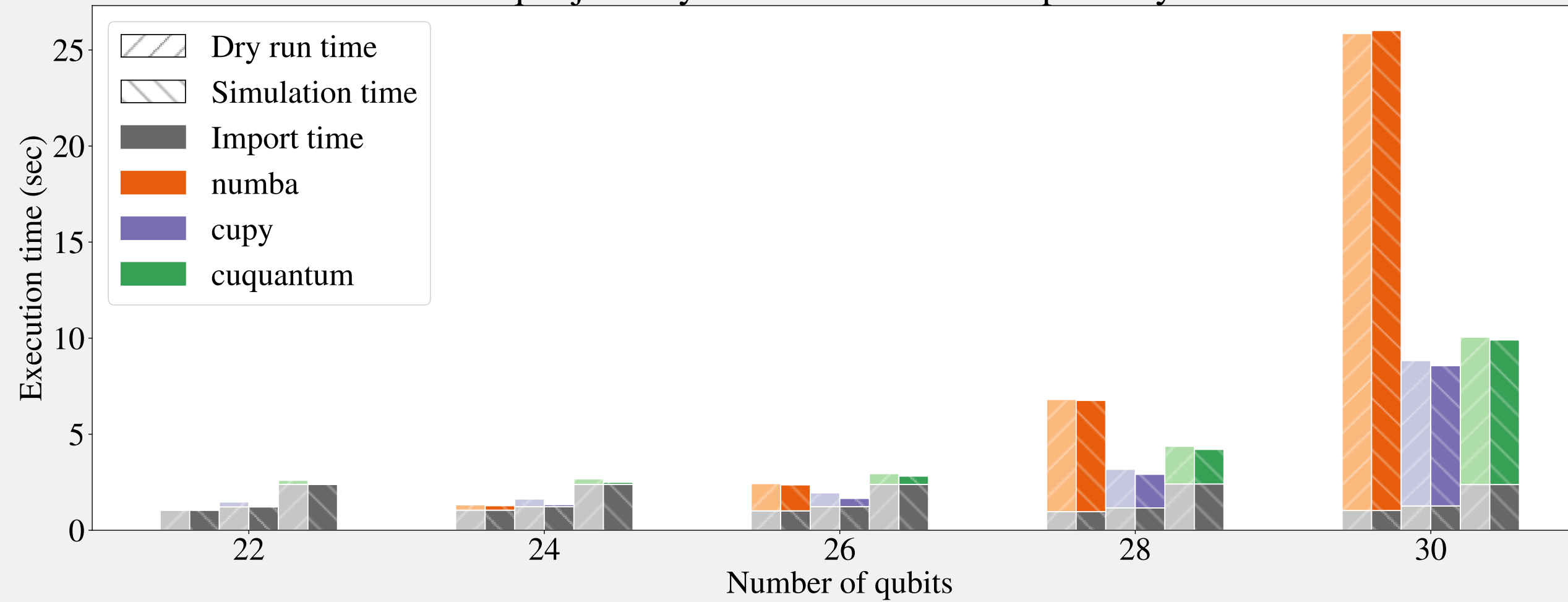
 [Documentation](#)

Thanks for listening!

Questions?

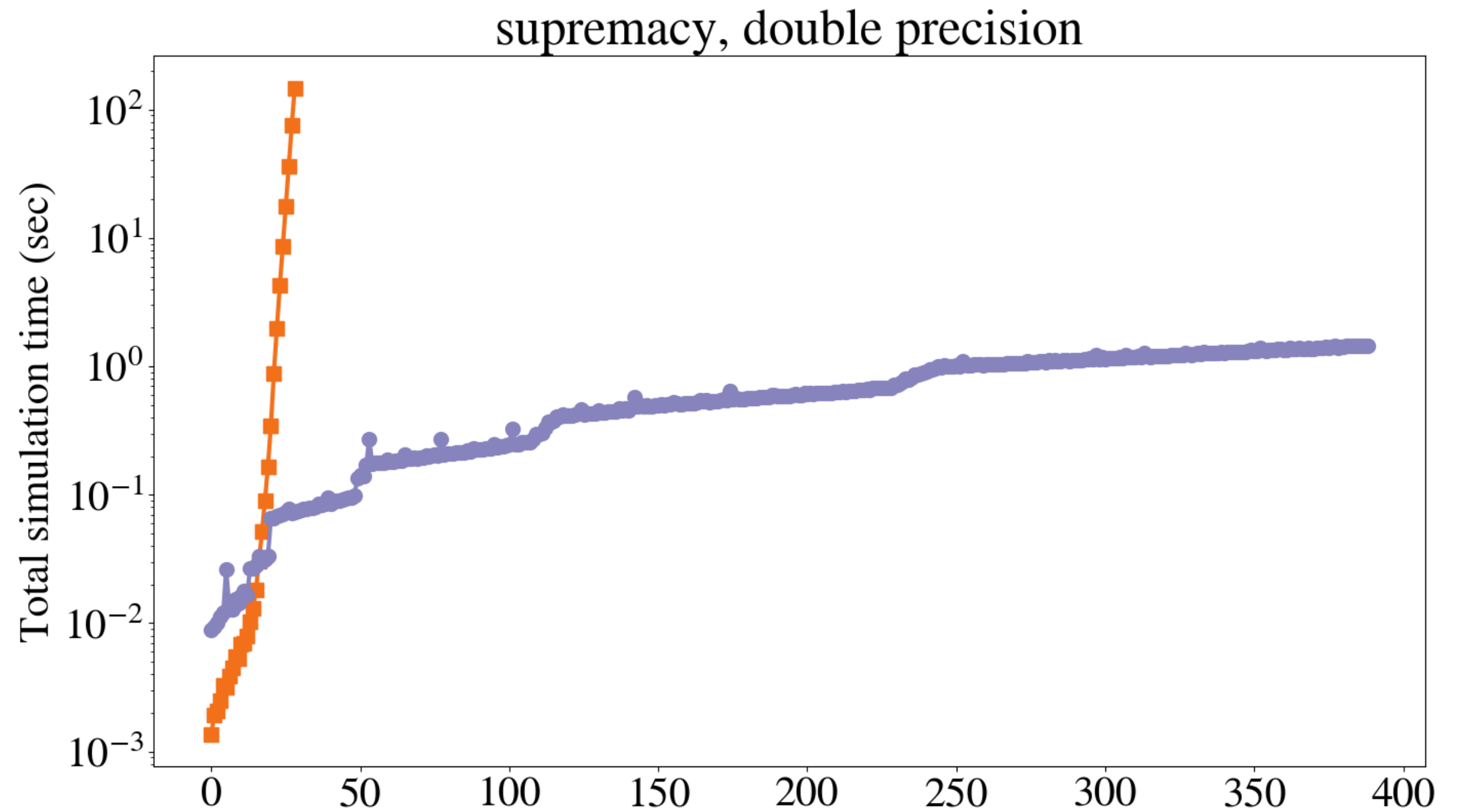
Backup slides

qibojit - Dry run vs simulation - supremacy



Comparison between import, dry run and simulation times for the three platforms of the qibojit backend.

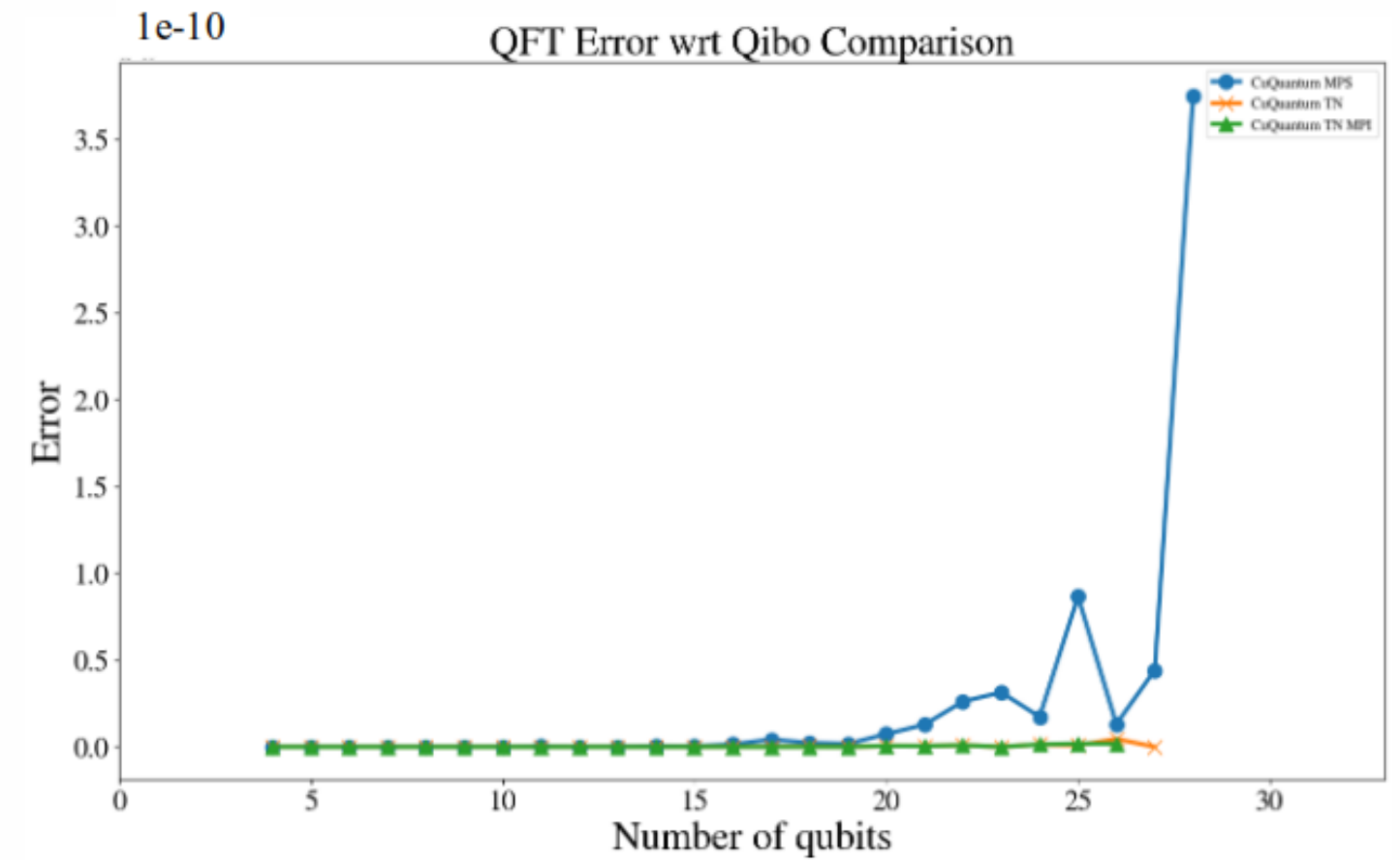
Running supremacy circuits
with 400 qubits using A100 GPU



Accuracy comparison: tensor network vs state vector

Tensor Network accuracy is generally high.

MPS tends to have lower accuracy but acceptable.



$$\Delta = \sum_{i=0}^{2^N} |\psi_{\text{Qibo}} - \psi_{\text{Qibotn}}|$$