

Benchmarking Variational Quantum Algorithms for the LUXE experiment

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LUXE



LUXE

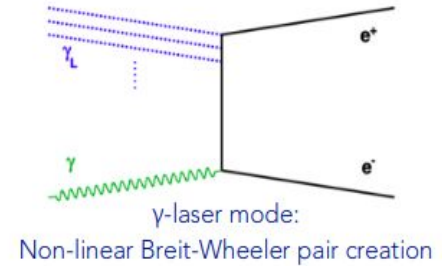
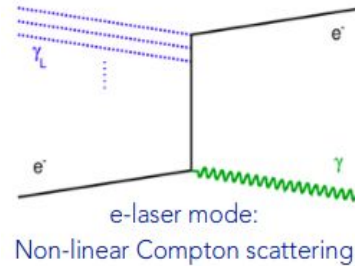
Laser und XFEL Experiment

Goal: Investigate the transition into the non-perturbative (strong-field) regime of QED

➔ Precision measurement of photon-photon, photon-electron interactions

How: Use high intensity laser (40-350 TW) and the european XFEL's electron beam (16.5 GeV)

Crucial: Measure number of positrons as a function of the laser intensity parameter $\xi = \sqrt{4\pi\alpha} \left(\frac{\varepsilon_L}{\omega_L m_e} \right) = \frac{m_e \varepsilon_L}{\omega_L \varepsilon_{cr}}$

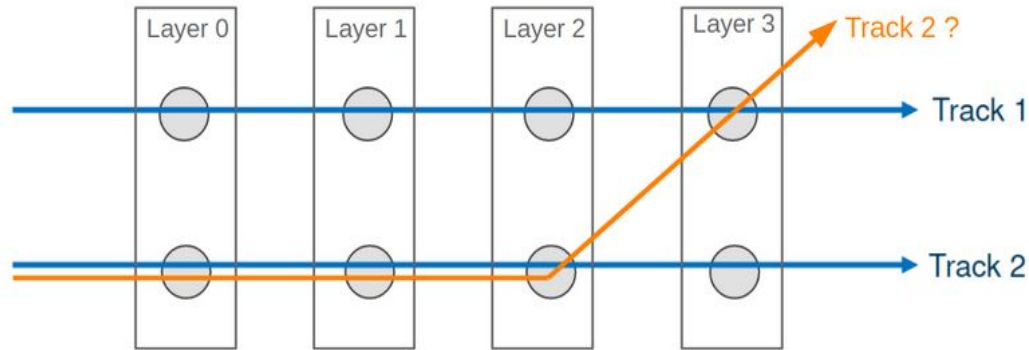


Track Reconstruction

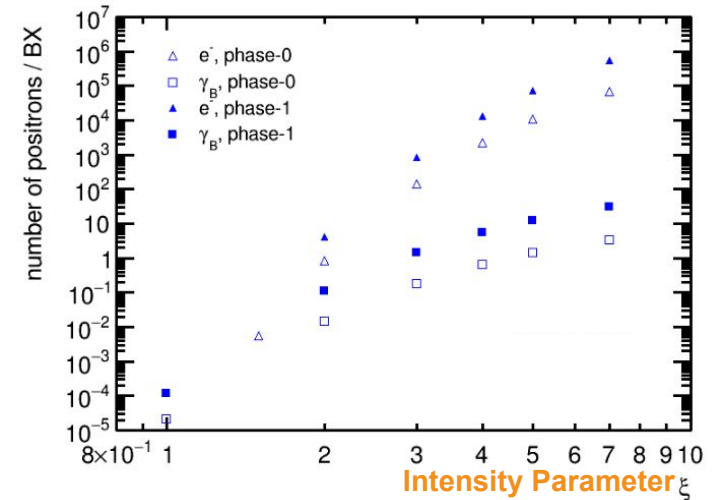
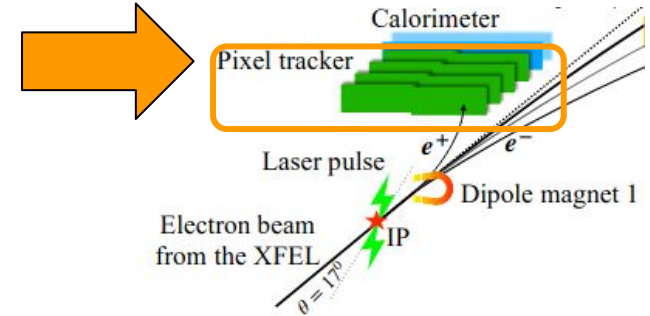
Positron Pixel Tracker

Positrons impinge on 4-layered binary hit/no-hit silicon detector (ALPIDE chip, 5×10^5 pixel of size $27 \times 29 \mu\text{m}^2$)

Challenge: Find tracks from a set of hits is computationally demanding



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Track Reconstruction

Novel approach: Particle track reconstruction using quantum computers

A Pattern Recognition Algorithm for Quantum Annealers

Frédéric Bapst, Wahid Bhimji, Paolo Calafiura, Heather Gray, Wim Lavrijsen, Lucy Linder ✉ & Alex Smith

Computing and Software for Big Science 4, Article number: 1 (2020) | [Cite this article](#)

1466 Accesses | 7 Citations | 14 Altmetric | [Metrics](#)

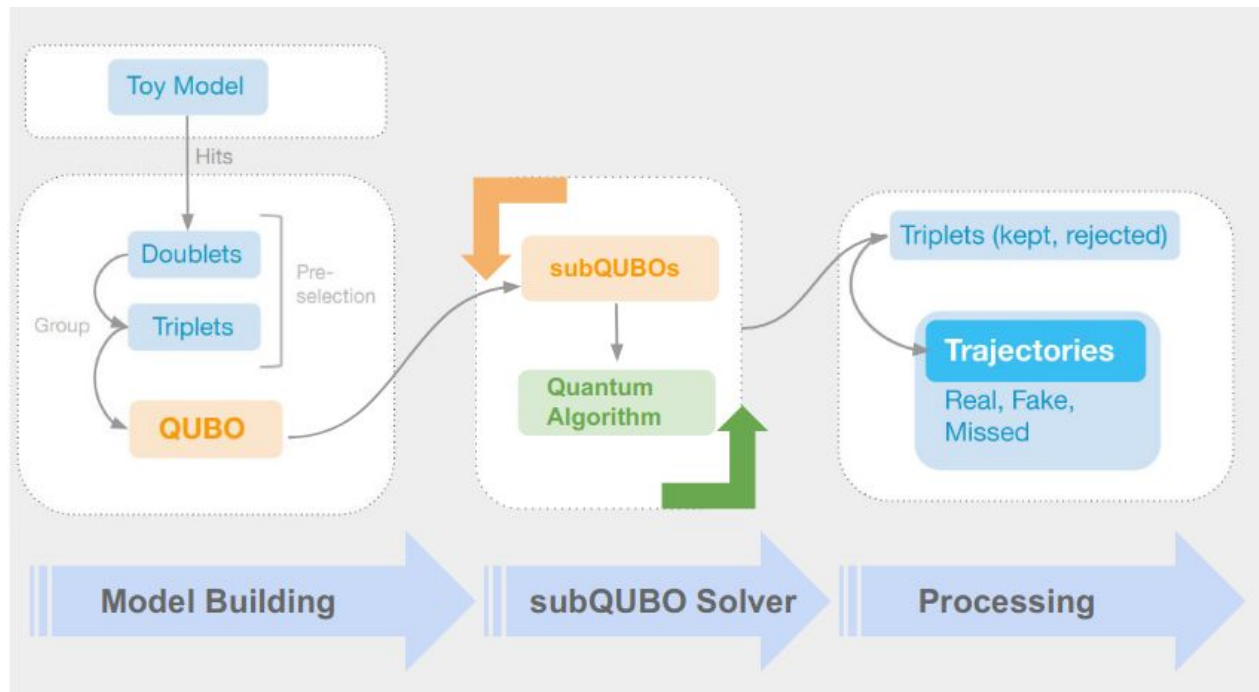
+ **Classical benchmark!**

➡ **QUBO (Quadratic Unconstrained Binary Optimization)**

The goal is to benchmark both on basis of:

1. Generated signal interactions at the IP
(T. G. Blackburn, A. J. MacLeod, B. King, [arXiv:2103.06673v2](#)),
2. Custom detector model where complexity can be controlled.

Quantum Algorithm Overview



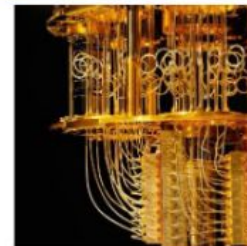
Devices used:

IBM Q

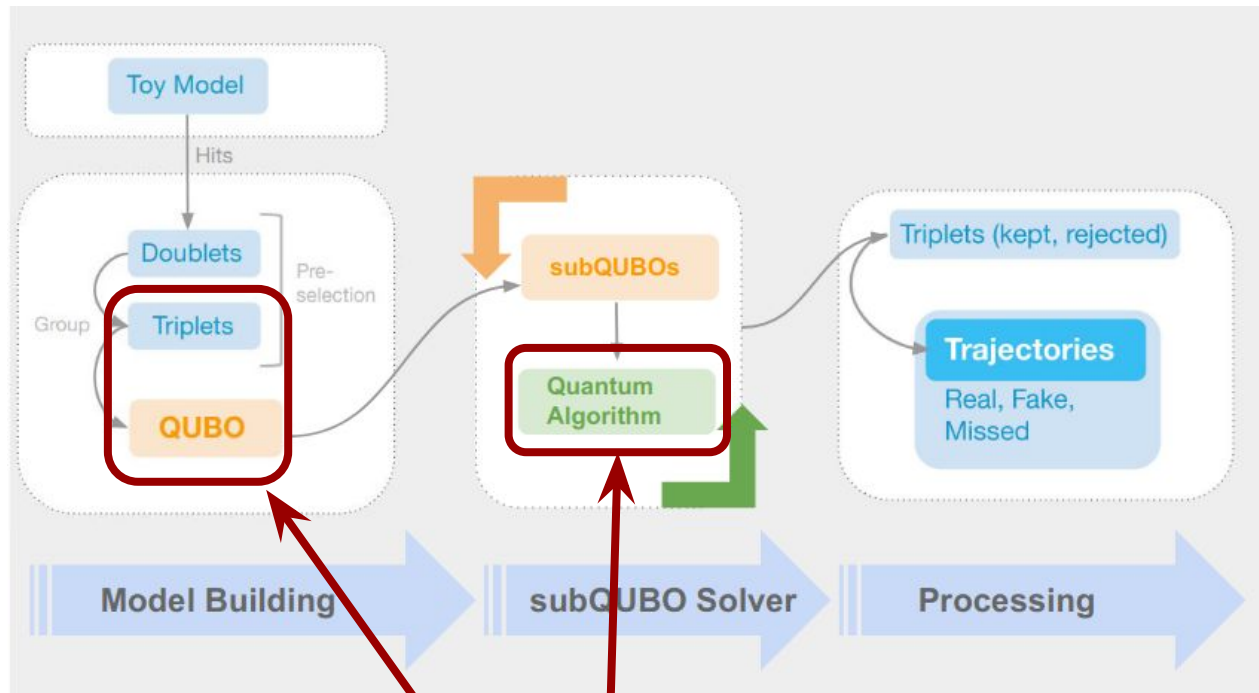
1-7 real qubit systems
available + qc simulators

IBM Quantum Experience:

(<https://quantum-computing.ibm.com/>)



Quantum Algorithm Overview

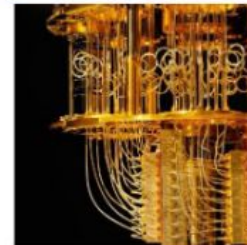


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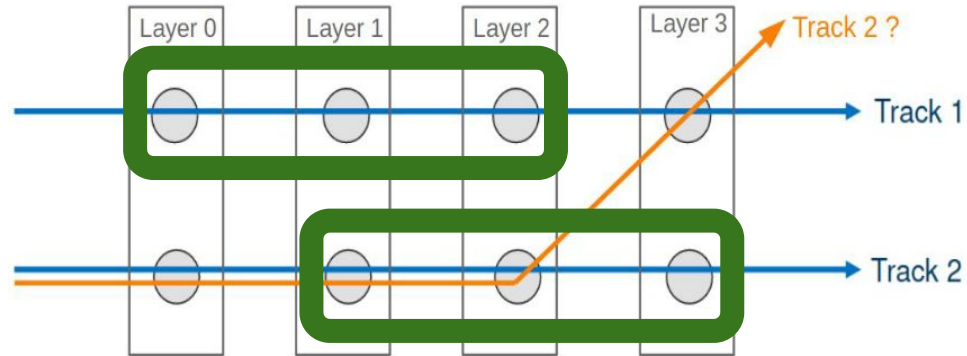
Tracking Formulation for a Quantum Computer

Grouping Hits

Task: Find formulation of the track reconstruction so that if simulated on a quantum computer, the ground state relates to correct tracks

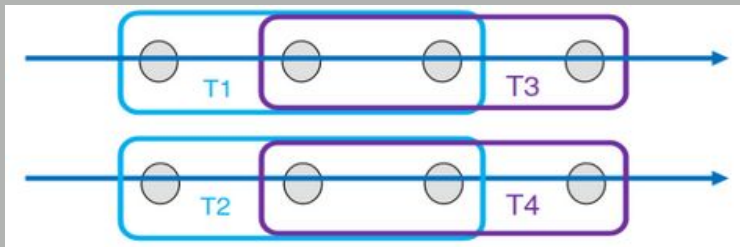
Step 1:

Form sets of triplets
(three consecutive hits)

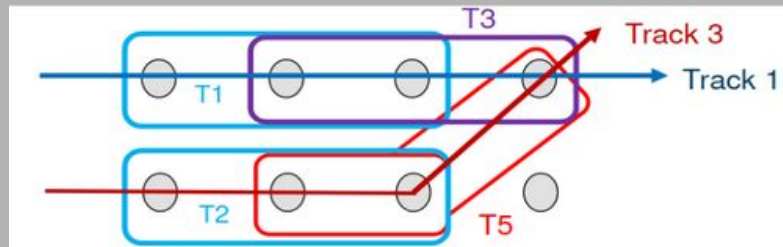


QUBO

Quadratic Unconstrained Binary Optimization



Well-aligned triplets



Conflict between triplets T3 and T5

Step 2: Maximize number of triplets that form well-aligned tracks and minimize triplets that contribute a conflict

- Express problem as a QUBO
- Minimizing the QUBO returns the best set of track candidates

QUBO

Quadratic Unconstrained Binary Optimization

In the QUBO formulation, triplets T_i are assigned a binary value, one if chosen and zero if discarded.

Closely related to Ising Model: Binary (0,1) instead of spins (-1,1).

$$O(a, b, T) = \sum_{i=1}^N a_i T_i + \sum_i^N \sum_{j < i}^N b_{ij} T_i T_j \quad T_i, T_j \in \{0, 1\}$$

quality of triplet
with
quality parameter a_i

Compatibility
between two
triplets

$$b_{ij} = \begin{cases} -S(T_i, T_j), & \text{if } (T_i, T_j) \text{ form a quadruplet,} \\ \zeta & \text{if } (T_i, T_j) \text{ are in conflict,} \\ 0 & \text{otherwise.} \end{cases}$$

Hybrid Algorithms

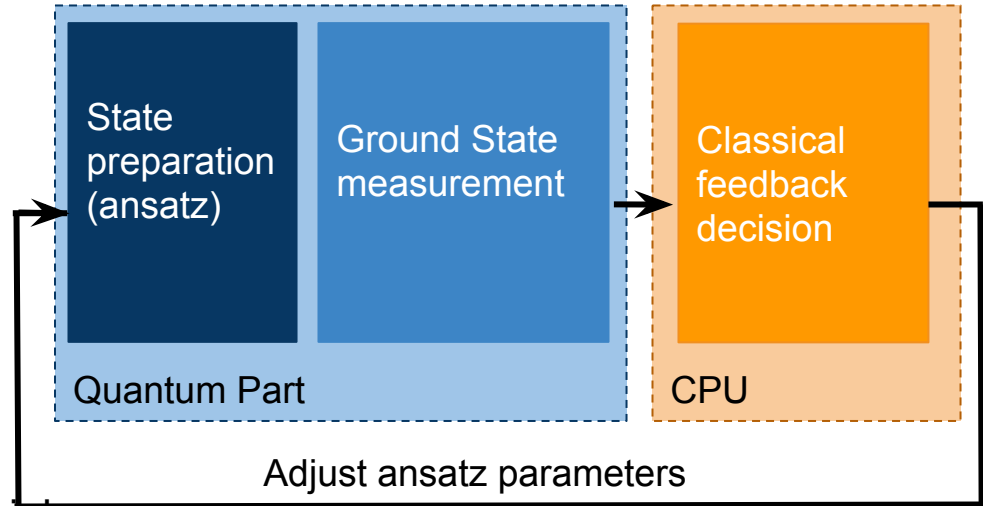
VQE/QAOA

Step 3: Minimize the QUBO

Hybrid: Quantum subroutine run inside of a classical optimization loop

Variational-Quantum-Eigensolver (VQE): hybrid algorithm used to find eigenvalues of a matrix

The **Quantum Approximate Optimization Algorithm (QAOA)** can be viewed as a special case of VQE

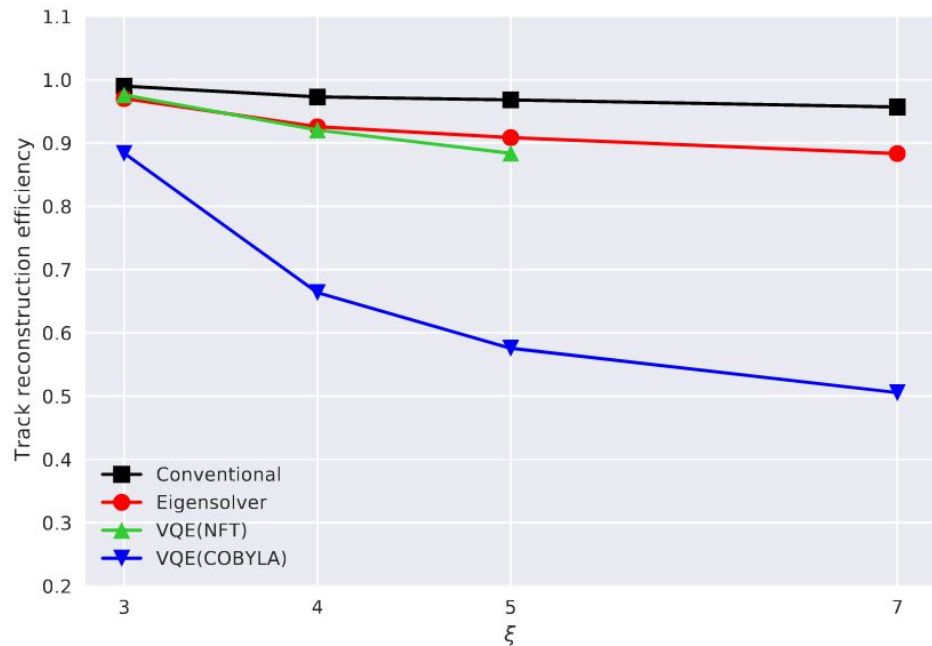


Track Reconstruction Efficiency

Benchmarking VQE(NFT) and VQE(COBYLA) on Simulator

- Two optimizers considered: NFT and COBYLA
- Eigensolver refers to solving subQUBO through diagonalizing of the Hamiltonian

VQE's performance depends on optimizer!

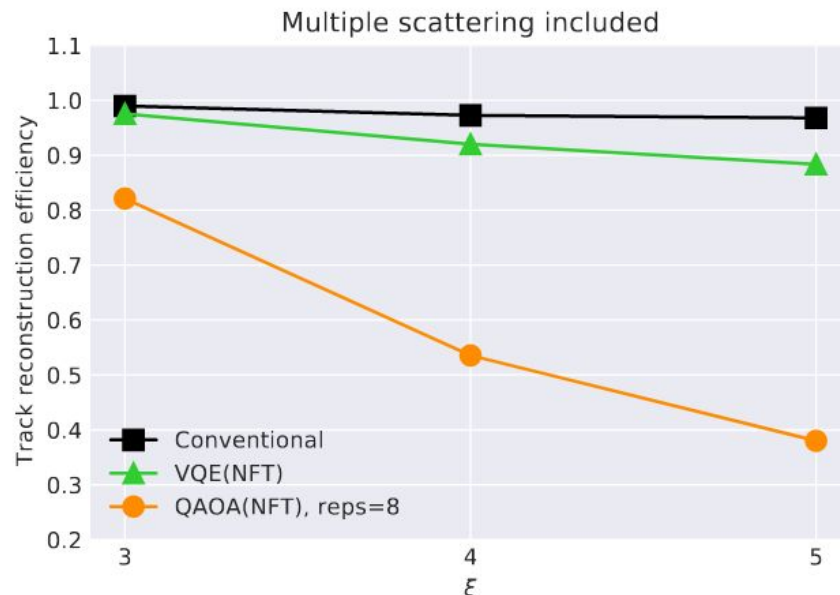


Track Reconstruction Efficiency

Benchmarking QAOA(NFT) and VQE(NFT) on Simulator

- QAOA's efficiency depends on the number of repeats (reps)
- Both are benchmarked using NFT optimizer

VQE outperforms QAOA using the NFT optimizer



Summary and Outlook

Goal: Study the limits for a quantum algorithm on **noisy device** to tackle particle track reconstruction for the LUXE experiment (positron tracker)

Here, VQE and QAOA were compared:

- VQE(NFT) depends on optimizer. NFT outperforms COBYLA
- VQE outperforms QAOA (8 reps)

Future outlook:

- Make use of real devices and develop error mitigation techniques
- Future study of solver's dependencies (reps, optimizer)
- Study impact of QUBO division method

Thank you

QAOA

Solving the subQubo

- QAOA can be viewed as a special case of VQE.
- Hamiltonian contains only Z terms, we do not need to change the basis for measurements.

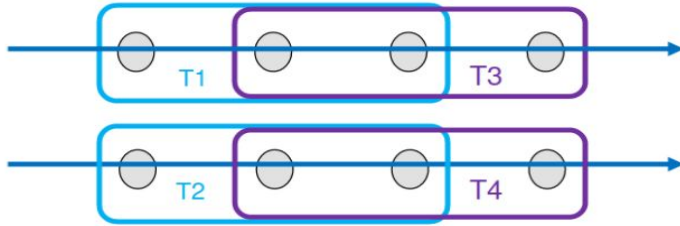
Differences to VQE:

- The form of the ansatz is limited
- Restricted to Ising Hamiltonians
- In QAOA our goal is to find the solution to the problem. To do that we don't need to find the ground state.

QUBOs

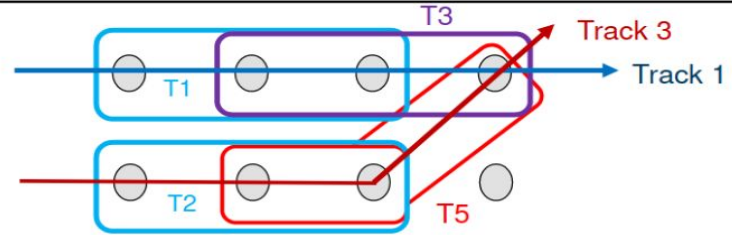
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[T1, T2, T3, T4] → combinations:

T1T2	T1T3	T1T4	T2T3	T2T4	T3T4
↓	↓	↓	↓	↓	↓
+0	-S	+0	+0	-S	+0



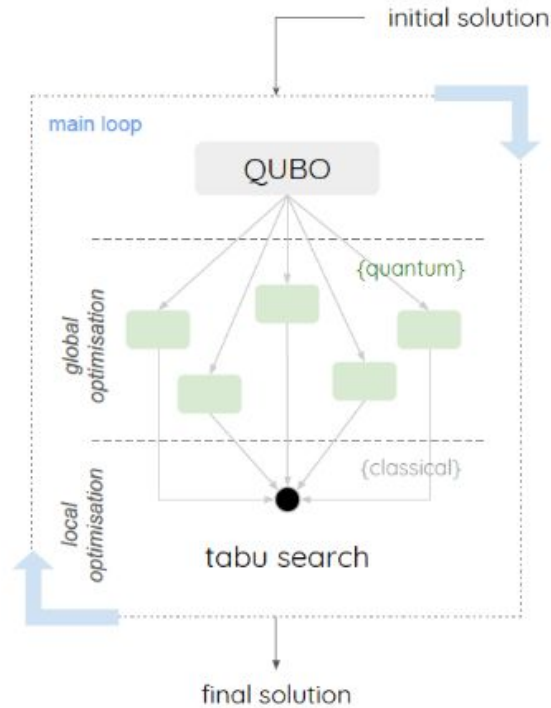
[T1, T2, T3, T5] → combinations:

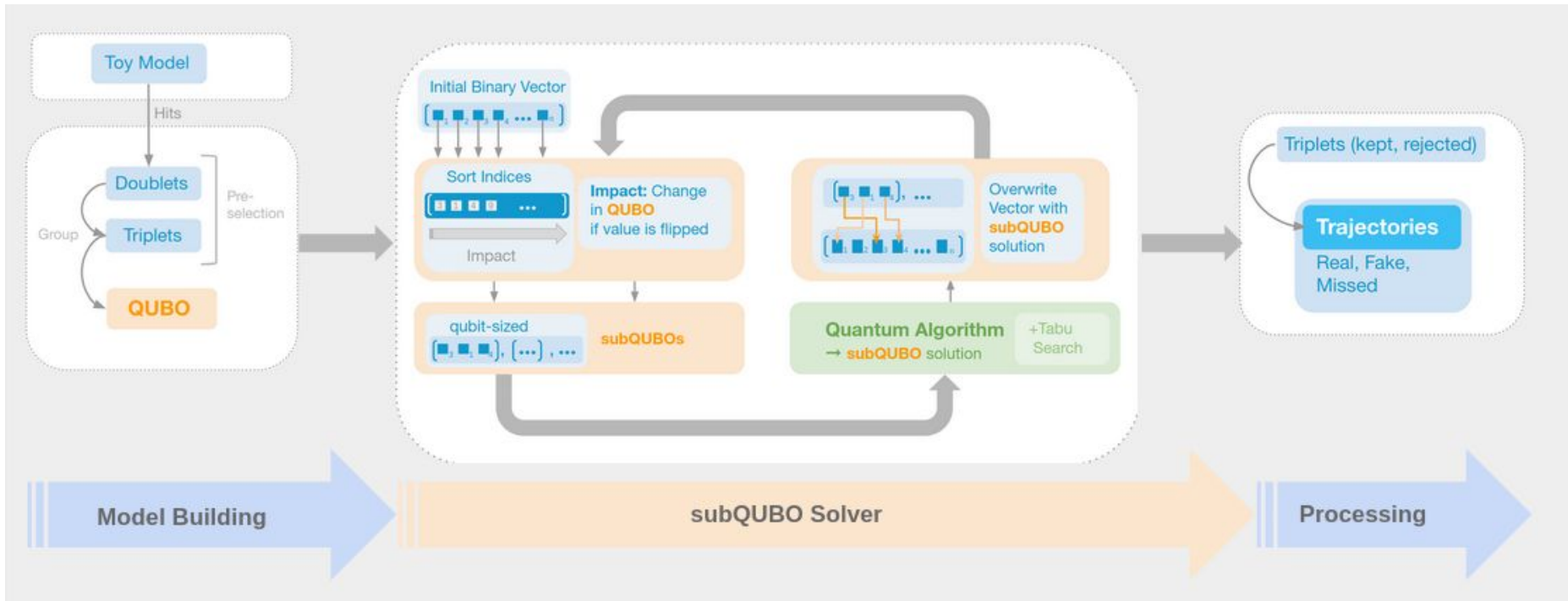
T1T2	T1T3	T1T5	T2T3	T2T5	T3T5
↓	↓	↓	↓	↓	↓
+0	-S	+0	+0	-S	conflict

SubQubos

Problem: Devices restricted to small number of qubits

- Big QUBOS cannot be simulated
- Break QUBO into subsets → subQUBOS!
- Iterated vector converges to solution vector





LUXE setup

