

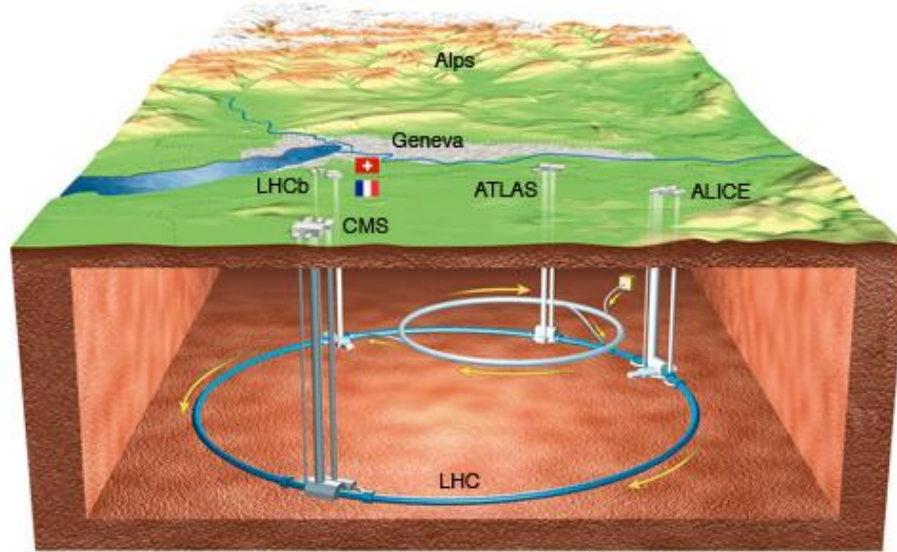
Quantum Computing for Track Reconstruction at LHCb

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The LHCb detector

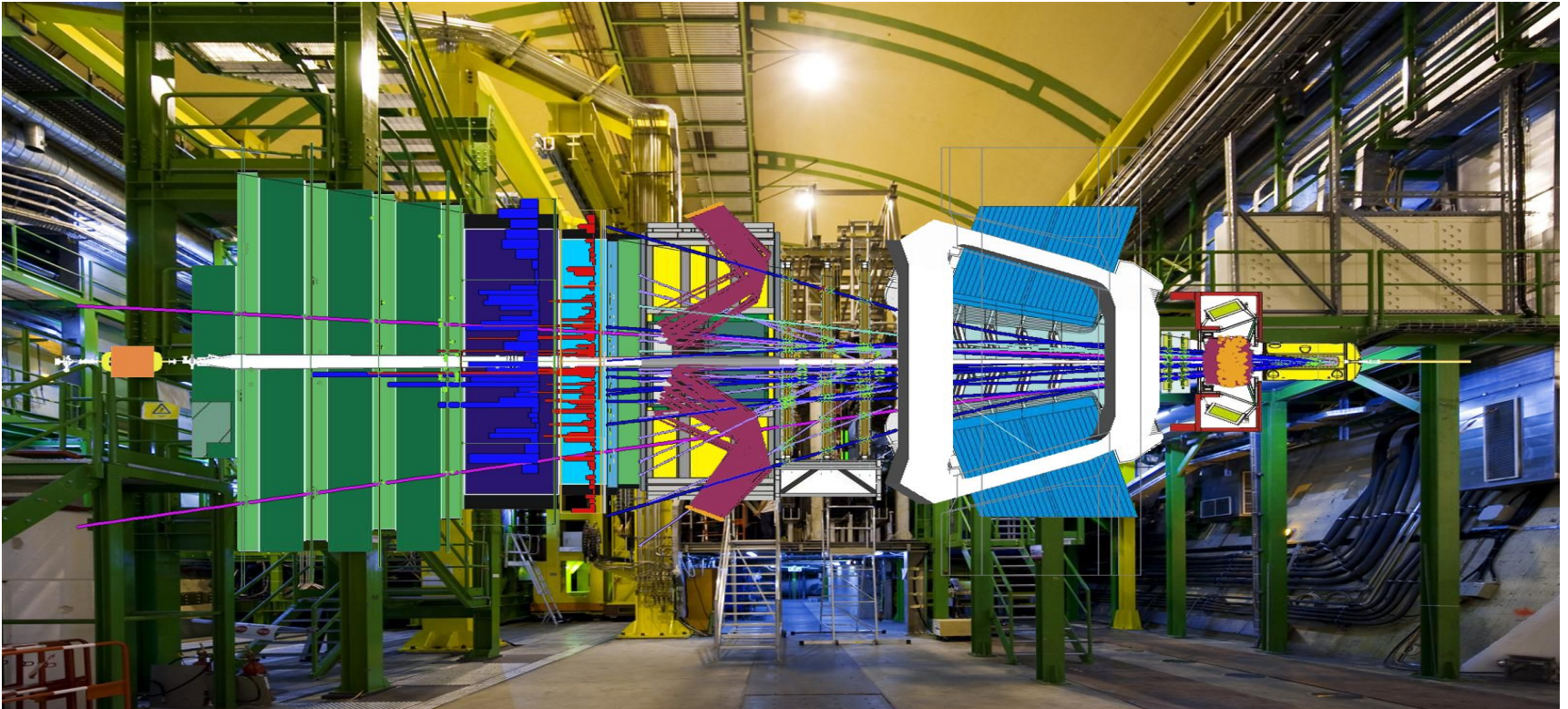
One of the 4 main experiments @ Large Hadron Collider at CERN



- Initially designed for the study of the **b,c-quarks**
- Now evolved into a general purpose spectrometer in the forward region
- Look for hints of BSM physics

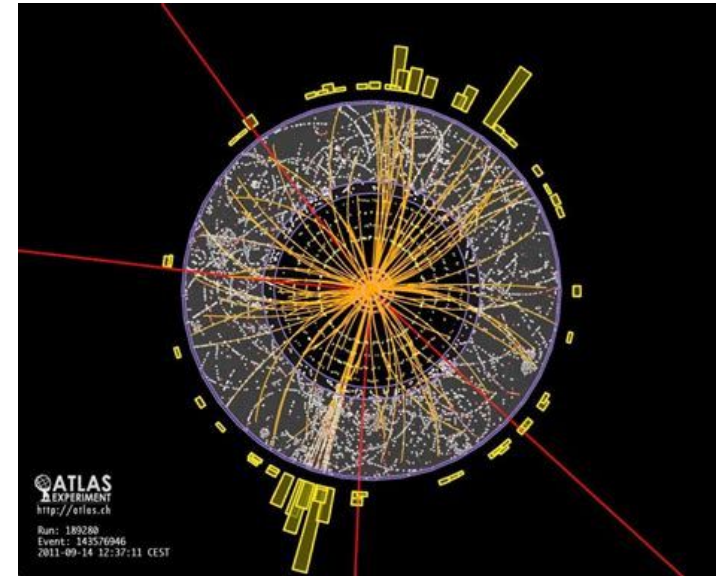
How does an event look like?

Reconstruct events 40 Million times per second.



Track reconstruction

- Recover the original trajectories from signals left by **charged particles**
 - signals \rightarrow 3D points or **hits**
 - need efficient distinction between the combinations of hits that are of interest and those that aren't
- Typical event: large number of **tracks**, modelled by a collection of **segments**

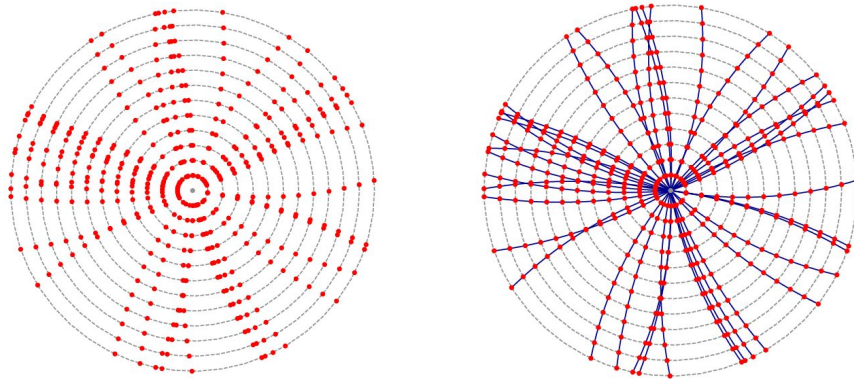


Track Reconstruction

- **Local tracking methods**: steps are performed sequentially. Some studies exist on QC for local tracking methods [arXiv:2104.11583]
- **Global tracking methods**: all hits are processed by the algorithm in the same way. Global algorithms are **clustering** algorithms. E.g.: QAOA, quantum annealing, Hopfield Networks, Hough transform

→ LHCb's current method of [search by triplet](#)

→ Focus of this talk: *global* algorithms



QC for Track Reconstruction

- Quantum Computing has very interesting prospects of improvements in algorithm **complexity/timing**
- This talk: two track reconstruction algorithms
- Define **Ising-like** $H^{\text{TrackReco}}(\text{hits})$:

$$H = -\frac{1}{2} \sum_{ij} \omega_{ij} \sigma_z^i \sigma_z^j - \sum_i \omega_i \sigma_z^i$$

→ $\mathbf{H}_{\min}^{\text{TrackReco}}$ == solution with the correct reconstructed tracks

QC for Track Reconstruction

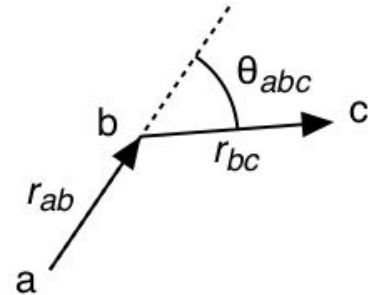
Ising-like Hamiltonian:

$$H = -\frac{1}{2} \sum_{ij} \omega_{ij} \sigma_z^i \sigma_z^j - \sum_i \omega_i \sigma_z^i$$

Segment $[S_{ab}]$: combination of hit **a** and hit **b**

→ in consecutive layers - for now

Hamiltonian accounts for **all** possible segments



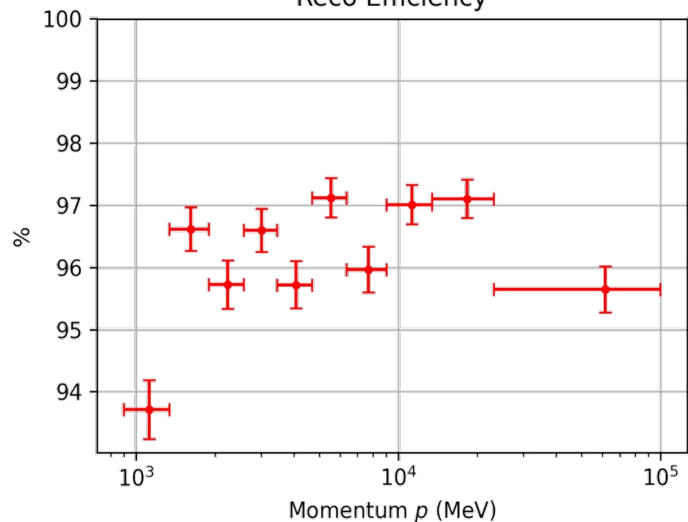
(Some) related results

[[JINST 18 \(2023\) 11, P11028](#)]

HHL algorithm

$$\nabla \mathcal{H} = 0 \Rightarrow AS = b$$

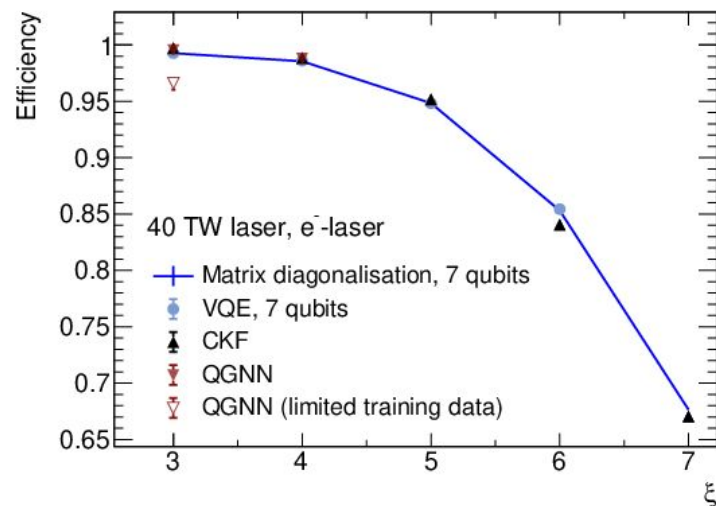
Reco Efficiency



[[Comput.Softw.Big Sci. 7 \(2023\) 1, 14](#)]

Variational Quantum Eigensolver

LUXE



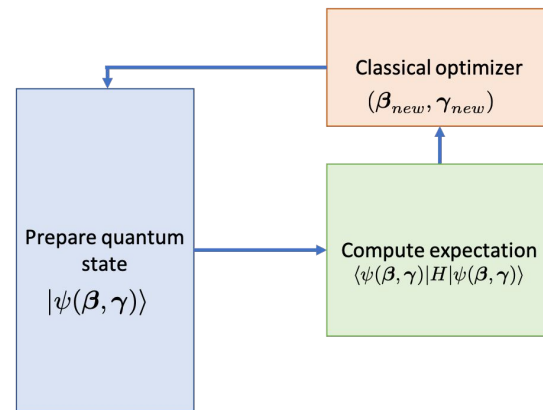
QAOA for Track Reconstruction

- Quantum Approximate Optimization Algorithm [[arXiv:1411.4028](#), [tutorial](#)]
- A **variational algorithm** ideal to solve combinatorial optimization problems, e.g. Max-Cut problem
 - ‘Finding an optimal object out of a finite set of objects’

$$|\psi(\beta, \gamma)\rangle = U(\beta)U(\gamma)...U(\beta)U(\gamma) |\psi_0\rangle$$

$$U(\beta) = e^{-i\beta H_B}, \quad U(\gamma) = e^{-i\gamma H_P}$$

- H_B : mixing Hamiltonian, H_P : **problem** Hamiltonian
- **Goal**: find optimal parameters $(\beta_{\text{opt}}, \gamma_{\text{opt}})$ such that the quantum state encodes the solution to the problem



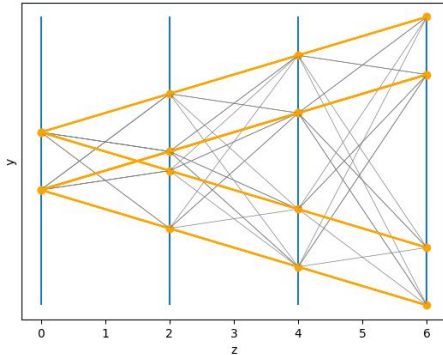
QAOA implementation

$$\mathcal{H} = -\frac{1}{2} \left[\underbrace{\left(\sum_{a,b,c} \frac{\cos^\lambda(\theta_{abc})}{r_{ab} + r_{bc}} s_{ab} s_{bc} \right)}_{(1)} - \alpha \underbrace{\left(\sum_{b \neq c} s_{ab} s_{ac} + \sum_{a \neq c} s_{ab} s_{cb} \right)}_{(2)} - \beta \underbrace{\left(\sum_{a,b} s_{ab} - N \right)^2}_{(3)} \right]$$

- (1) main term: favours aligned, short segments
- (2) 1st penalty term: forbids segments that share head/tail from belonging to the same track
- (3) 2nd penalty term: keeps the number of active segments equal to #hits

Results from simulation

- **Successful** implementation and validation for small simulations
- Scalability poses an issue, affecting especially the simulator
 - triplets instead of doublets → worse scalability

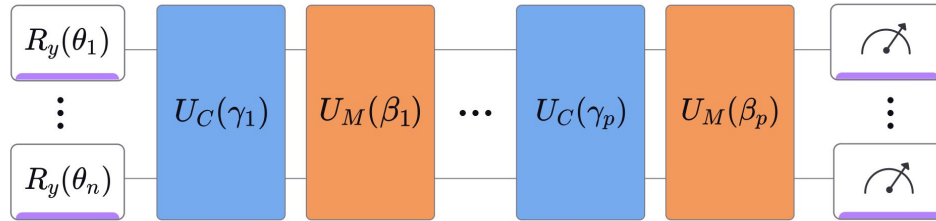


# tracks	# layers	#qubits (segments)	Circuit depth
2	3	8	103
2	4	12	223
3	3	18	497
3	4	27	1105
4	3	32	1553
4	4	48	3473
5	3	50	3775
5	4	75	8463

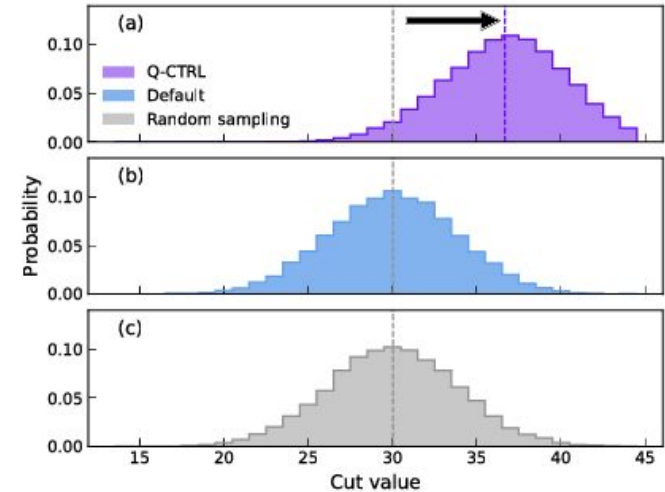
Modified QAOA (with P. Pariente, V. Chobanova, IFIC-UdC)

Using results from Q-CTRL and IBM

[[arXiv:2406.01743v1](https://arxiv.org/abs/2406.01743v1)]



→ pdf of finding the correct solution seem to decrease



Modified QAOA (with P. Pariente, V. Chobanova, IFIC-UdC)

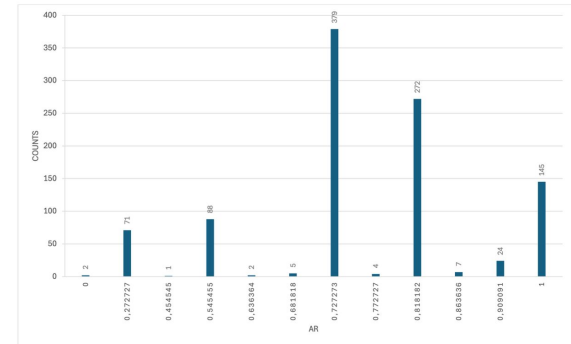
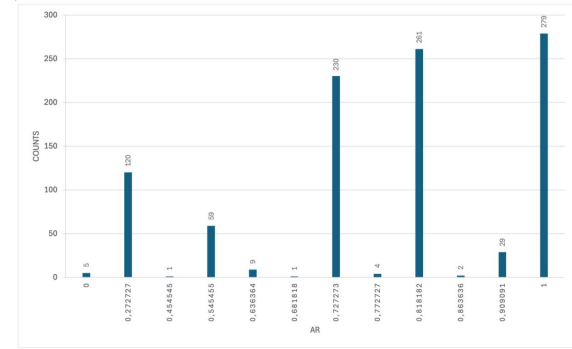
Approximation Ratio

$$AR(\vec{x}) = \frac{C(\vec{x}) - C_{max}}{C_{min} - C_{max}}$$

Success Probability

$$SP = \frac{\text{Nr. Optimal solutions}}{\text{Nr. shots}}$$

- ★ Higher depth \rightarrow higher clustering around $AR > 0.7$ for standard and modified QAOA
- ★ Modified QAOA has less occurrences with low AR, but also less at the exact solution



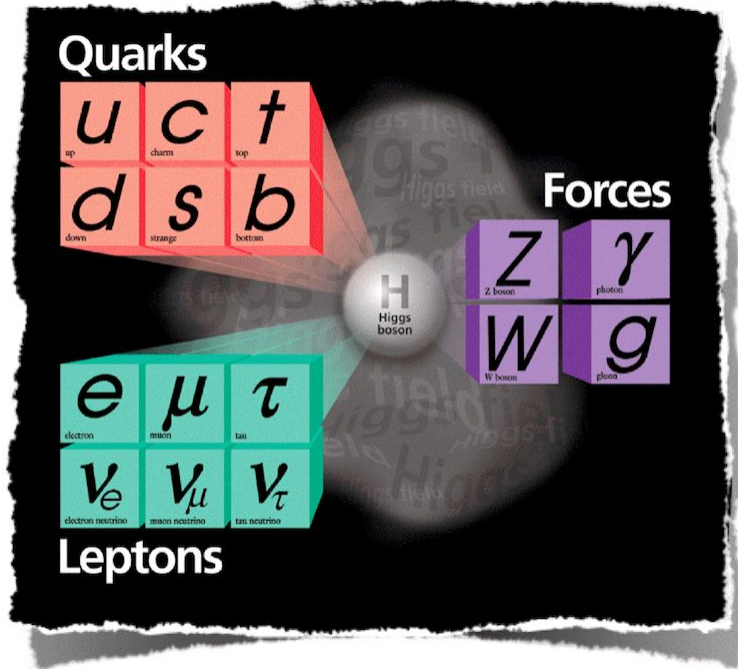
Ongoing/future work

- Sustainability & Quantum: project ongoing with **GSoC & IFIC-UV**
 - Using [ACTS](#) as framework
 - Different technologies are being considered
 - Attention needed for a comparison as fair as possible between Quantum and Classical
- Try simulation using **Rydberg atoms**
- **Distributed** QAOA (OakRidge)
- Results for tracking with QAOA at LHCb being currently **summarized in a paper**
- Further applications of QAOA for HEP with better scalability and/or different use-cases

Thanks for your attention!

The Standard Model of Particle Physics

A successful theory that describes the interactions among particles ...



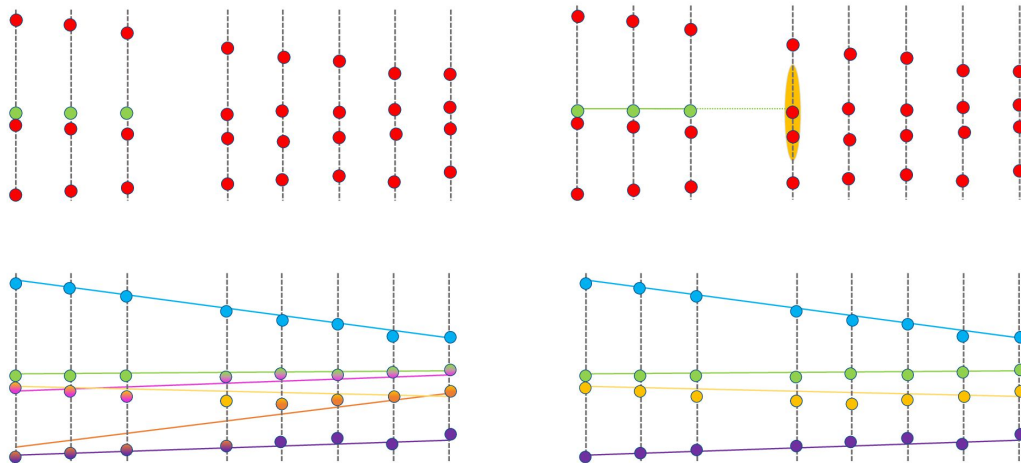
... but fails to explain several phenomena observed in the Universe:

- Neutrinos masses
- Origin of Dark Matter & Dark Energy
- etc

⇒ need of **Beyond the Standard Model physics!!**

Local tracking methods [[arXiv:2104.11583](https://arxiv.org/abs/2104.11583)]

1. Seeding
2. Track building
3. Cleaning
4. Selection



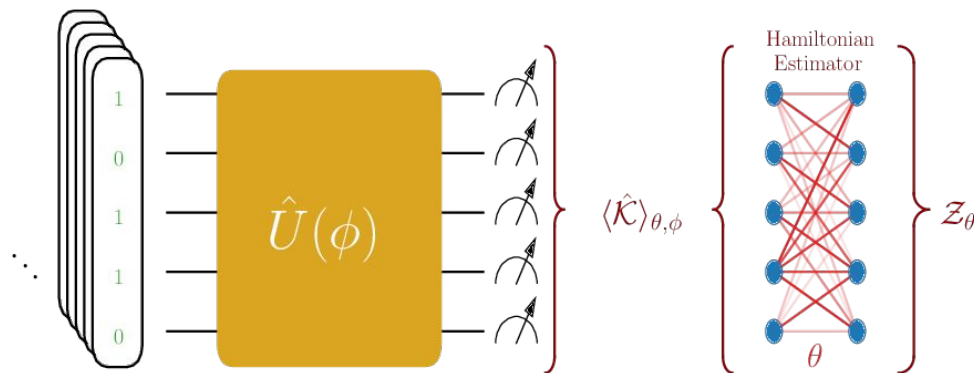
Tracking stages	Input size	Output size	Classical complexity	Quantum complexity
Seeding	$O(n)$	k_{seed}	$O(n^c)$ (Theorem 2)	$\tilde{O}(\sqrt{k_{\text{seed}} \cdot n^c})$ (Theorem 3)
Track Building	$k_{\text{seed}} + O(n)$	k_{cand}	$O(k_{\text{seed}} \cdot n)$ (Theorem 4)	$\tilde{O}(k_{\text{seed}} \cdot \sqrt{n})$ (Theorem 5)
Cleaning (original)	k_{cand}	$O(k_{\text{cand}})$	$O(k_{\text{cand}}^2)$ (Theorem 6)	—
Cleaning (improved)	k_{cand}	$O(k_{\text{cand}})$	$\tilde{O}(k_{\text{cand}})$ (Theorem 7)	—
Selection	$O(k_{\text{cand}})$	$O(k_{\text{cand}})$	$O(k_{\text{cand}})$ (Theorem 8)	—
Full Reconstruction	n	$O(n^c)$	$O(n^{c+1})$ (Theorems 2, 4, 7, 8)	$\tilde{O}(n^{c+0.5})$ (Theorems 3, 5, 7, 8)
Full Reconstruction with $O(n)$ reconstructed tracks	n	$O(n)$	$O(n^{c+1})$ (Theorems 2, 4, 7, 8)	$\tilde{O}(n^{(c+3)/2})$ (Theorem 9)

n : number of particles, c : number of hits, k_{seed} : total number of generated seeds, k_{cand} : number of track candidates

Another possible idea

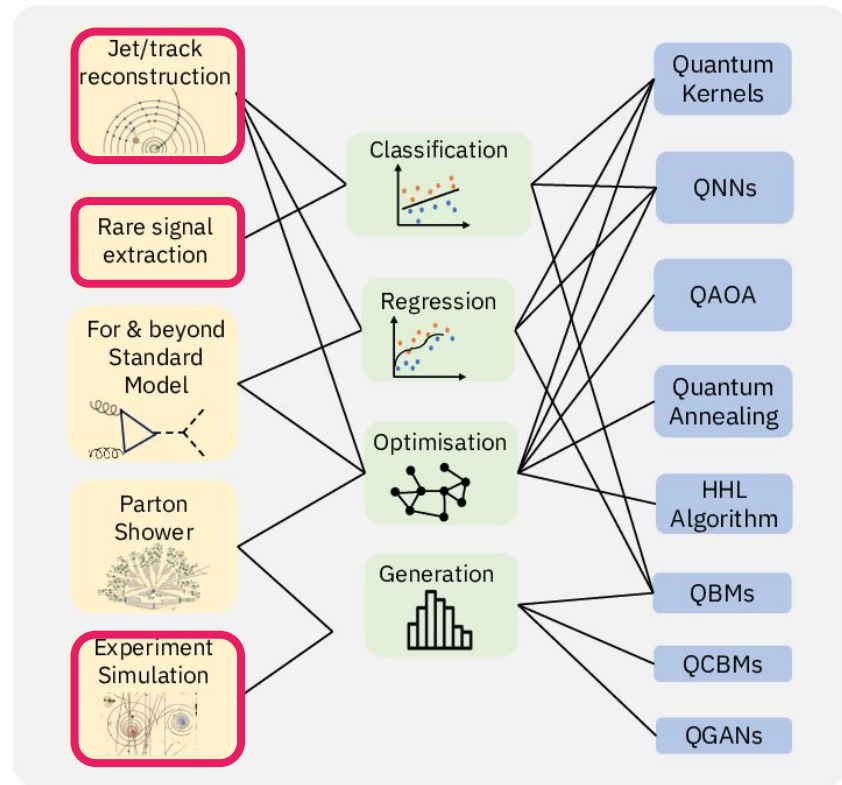
‘Quantum-probabilistic Hamiltonian learning for generative modelling & anomaly detection’ [[arXiv:2211.00380v2](https://arxiv.org/abs/2211.00380)]

- Using LHC data & following a Quantum Hamiltonian-Based Models (QHBM) approach
- Generative modelling
- Anomaly detection



HEP use-cases

- [Summary of the QC4HEP WG](#)
- Focused mostly in projects concerning experimental particle physics at **LHC** and **LHCb**
- Events are **quantum** in nature, but measurements are **classical**
- QC4HEP WG continues to meet **periodically + 1 in-person meeting per year**, open to new groups joining



Recent progress QC4HEP-ex

iHVA (C. Tüysüz et al., DESY)

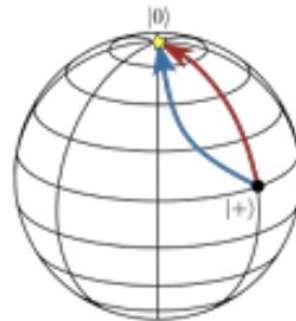
- QITE-inspired
- Avoid Barren Plateaus from QAOA
- Not unique set of gates possible
- Geodesics for parametrized quantum circuits also considered by [people at IFIC](#)

[[arXiv:2408.09083](#), [Presentation at QC4HEP](#)]

Target Hamiltonian: $-Z$

$$\text{QAOA} \quad |+\rangle \xrightarrow{R_Z(\theta)} \xrightarrow{R_X(\theta)} |\phi_R(\theta)\rangle$$

$$\text{iHVA} \quad |+\rangle \xrightarrow{R_Y(\theta)} |\phi_I(\theta)\rangle$$



The **iHVA** follows the geodesic.
This leads to faster convergence.