

Investigating the use of Quantum Annealing for Tracking, using the TrackML Dataset.

TrackML Challenge Grand Finale
CERN, July 1-2 2019

Jean-Roch Vlimant
on behalf of the **HEP.QPR**
and **QMLQCF** project teams,
with material from Heather M. Gray.

Motivation



Charged particle tracking is a computing intensive task, and mostly combinatorics.

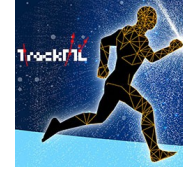
Embrace complexity and aim to resolve combinatorics with the use of quantum computing.

Familiarize with quantum computers, and start approaching problems differently.

Outline



- ♦ Introduction to D-Wave and quantum annealing (QA)
- ♦ Tracking with Segment classification as a QUBO problem
- ♦ Tracking with Triplet classification as a QUBO problem
- ♦ Ambiguity resolution with QA



The D-Wave Computing System

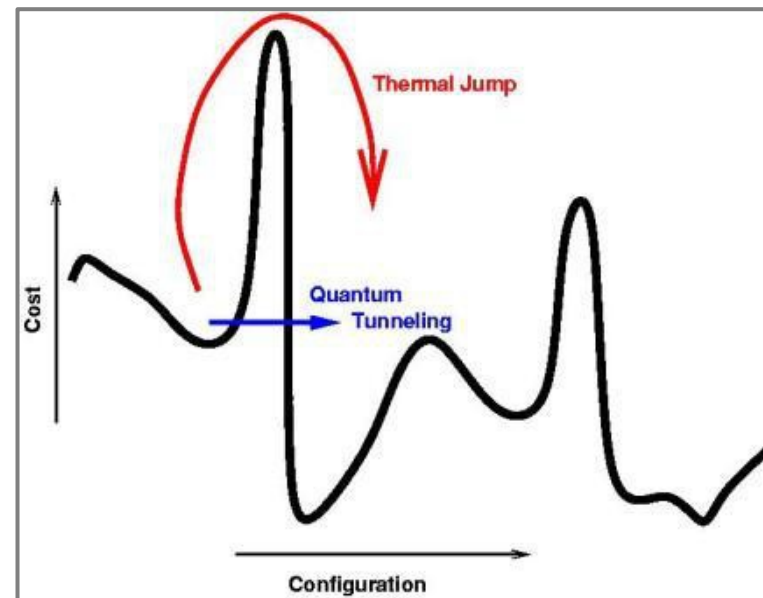
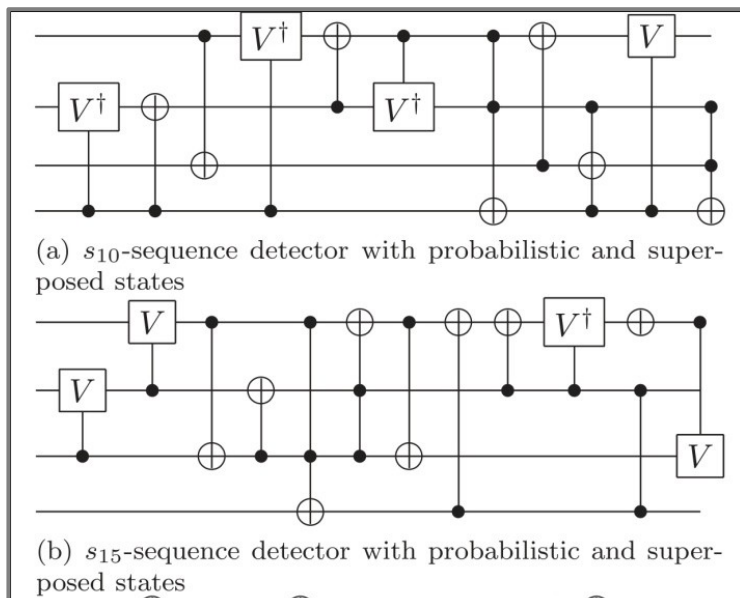
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TrackML Grand Finale, QA-Tracking, J-R Vlimant

D-Wave 2X™



qubit and qubit



Quantum Circuits

Series of quantum gates operating on a set of quantum states.

Quantum Annealing

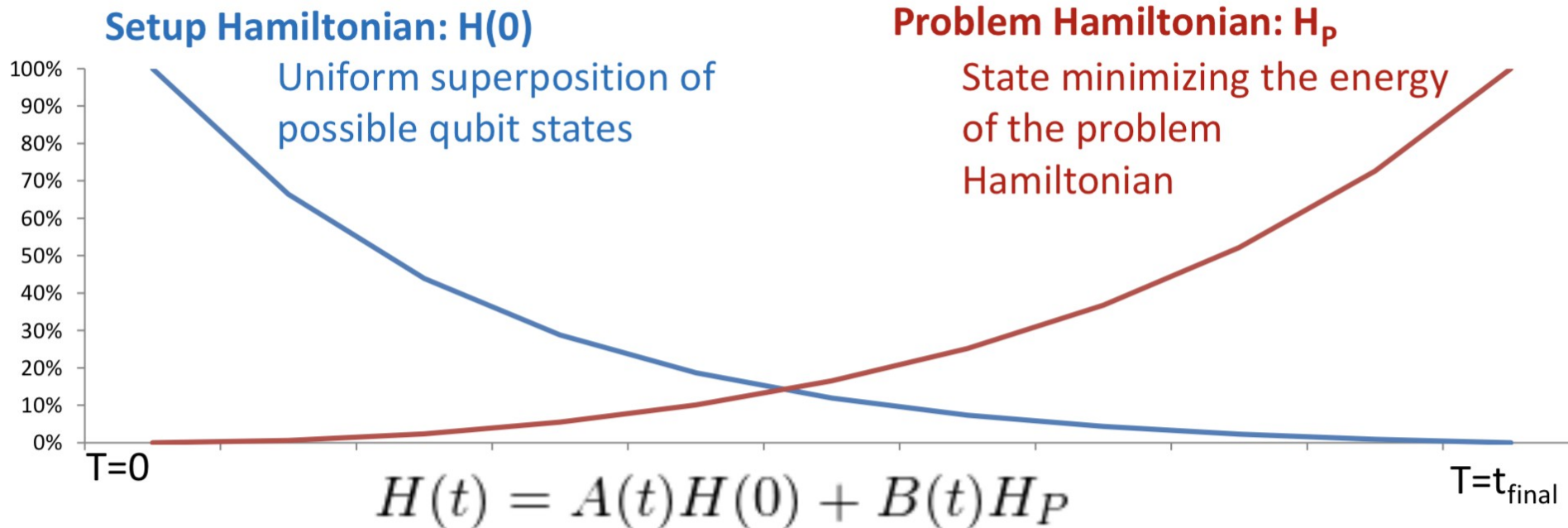
Evolution of a quantum system to a low T Gibbs state
That's D-Wave !



Quantum Annealing

Quantum Annealing

- System setup with trivial hamiltonian $H(0)$ and ground state
- Evolve adiabatically the hamiltonian towards the desired Hamiltonian H_p
- **Adiabatic theorem** : with a slow evolution of the system, the state stays in the ground state.



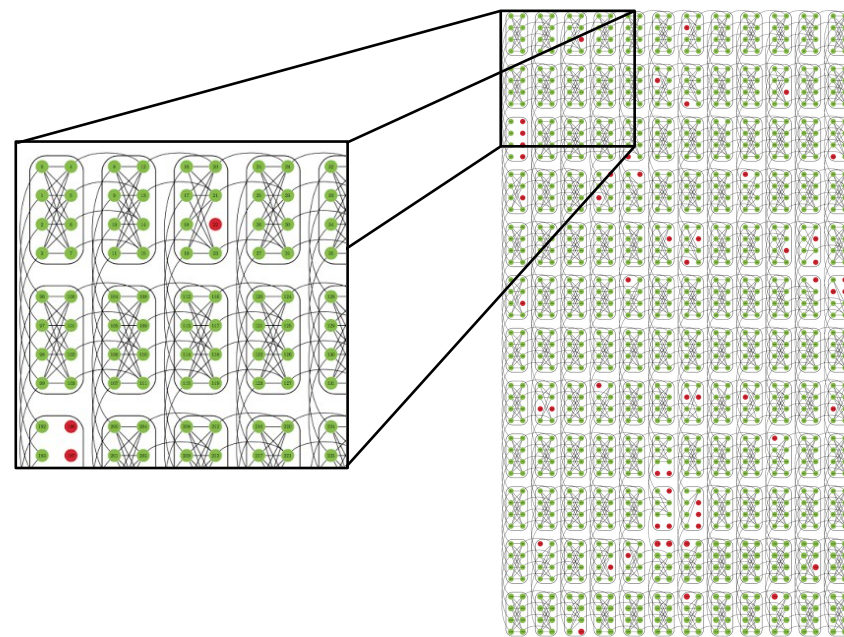
<https://arxiv.org/abs/quant-ph/0001106>
<https://arxiv.org/abs/quant-ph/0104129>

Space of Hamiltonian

$$H_{\text{Ising}} = \sum_i h_i \sigma_i^z + \sum_{ij} J_{ij} \sigma_i^z \sigma_j^z$$

Runs over all qubit pairs

External magnetic field Interactions



$$H_{\text{Ising}} = \sum_i h_i \sigma_i^z + \sum_{ij} J_{ij} \sigma_i^z \sigma_j^z$$

Runs over adjacent qubits

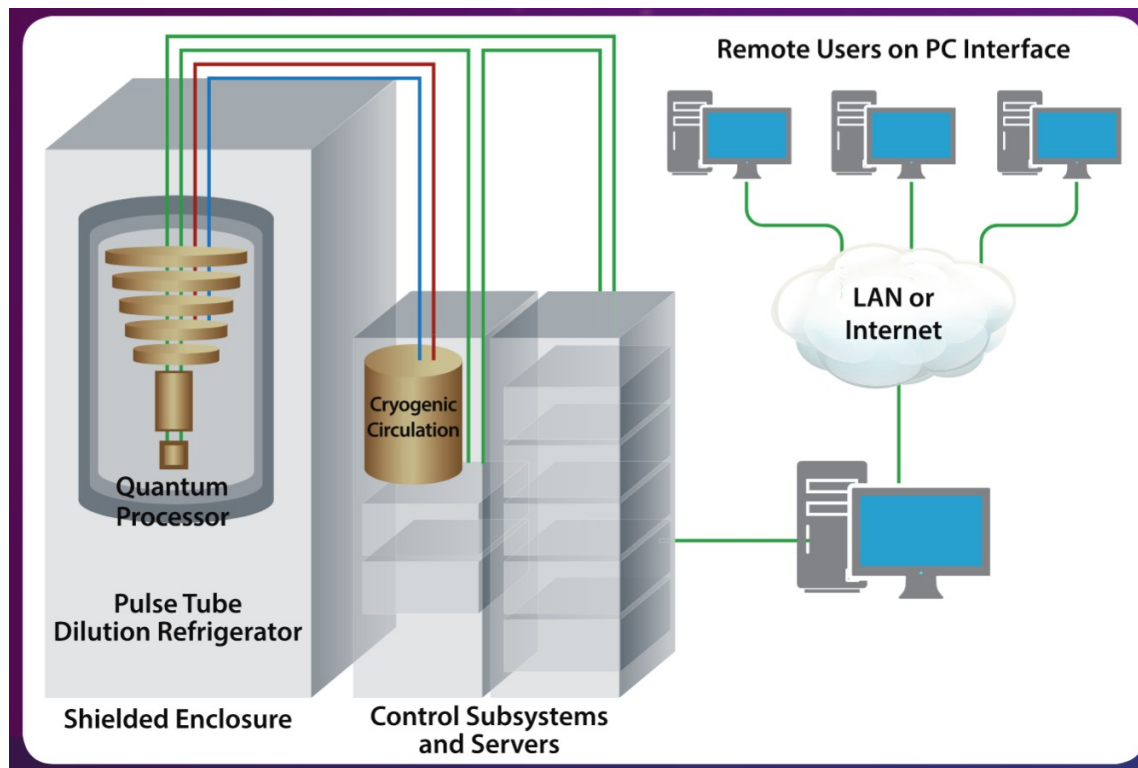
External magnetic field Interactions

- ~1000 qubits on chimera graph can only encode ~40 qubits full Ising Hamiltonian
- **Quadratic Unconstrained Binary Optimization (QUBO)** can be mapped to an Ising Hamiltonian with change of variable $\{0,1\} \leftrightarrow \{-1,1\}$
- Next generation (pegasus) will have better connectivity



Solving QUBO

Working on a D-Wave



- Web Interface to post the problem settings (Hp).
- A/synchronous processing.
- Distributed library for performing embedding
- Equivalent restapi to submit and retrieve solutions
- D-Wave processor as a service

<https://www.dwavesys.com/take-leap>

D-Wave Solutions



User provides
(\mathbf{h}_i , \mathbf{J}_{ij})

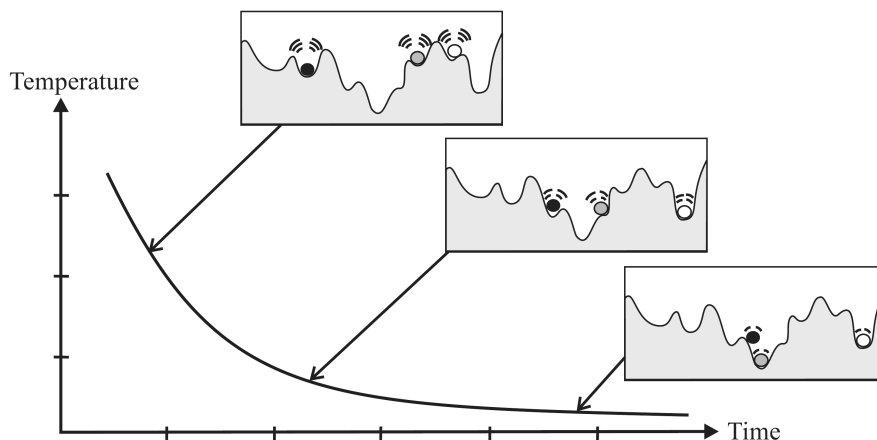
Multiple 5-20 μs
annealing cycles

D-Wave provides sampling
from lowest energy levels
(approx. Gibbs)
($\{\sigma_i\}_k$, $\mathbf{N}(\{\sigma_i\}_k)$)

Simulated Annealing



- Purely “classical” Monte-Carlo based method to find ground state of energy functions
- Random walk across phase space
 - accepting descent
 - accepting ascent with probability $e^{-\Delta E/kT}$
- Decrease T with time



Applied to the QUBO problem, and finds the **ground state.**

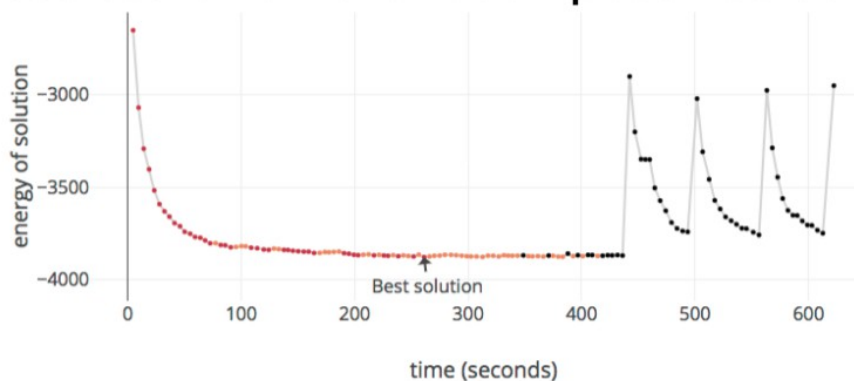
qbsolve

iterative hybrid classical/quantum algorithm

QBSOLV

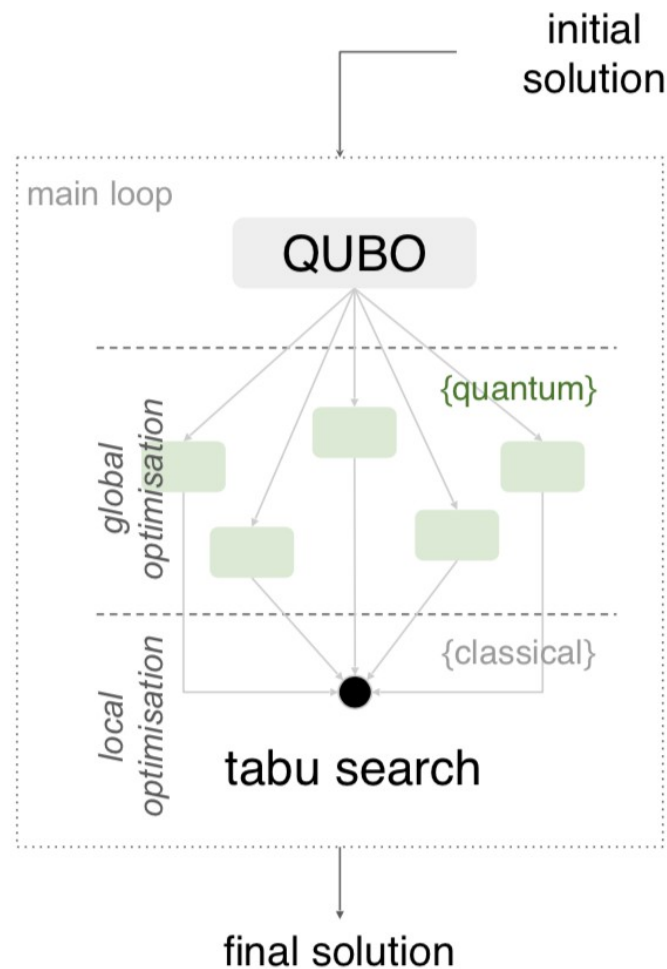
large and/or densely connected QUBOs split into sub-QUBOs fitting the D-Wave hardware.

Tabu search on the recomposed solutions.

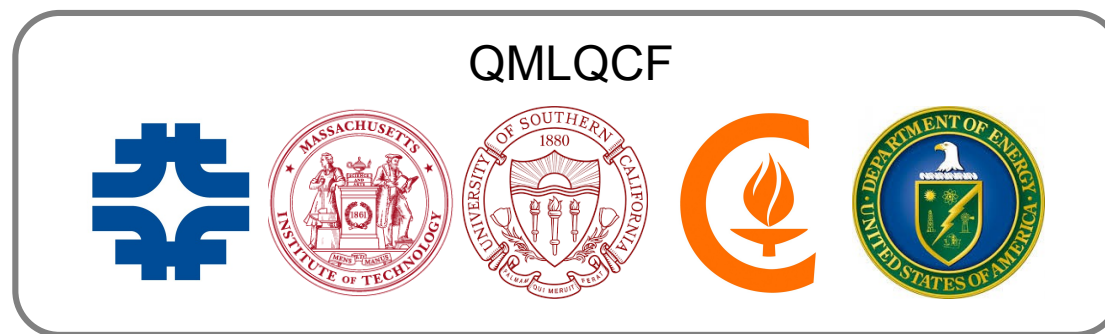


evolution of the solution in each [qbsolv](#) loop.

The solution is sometimes randomised to escape local minima.



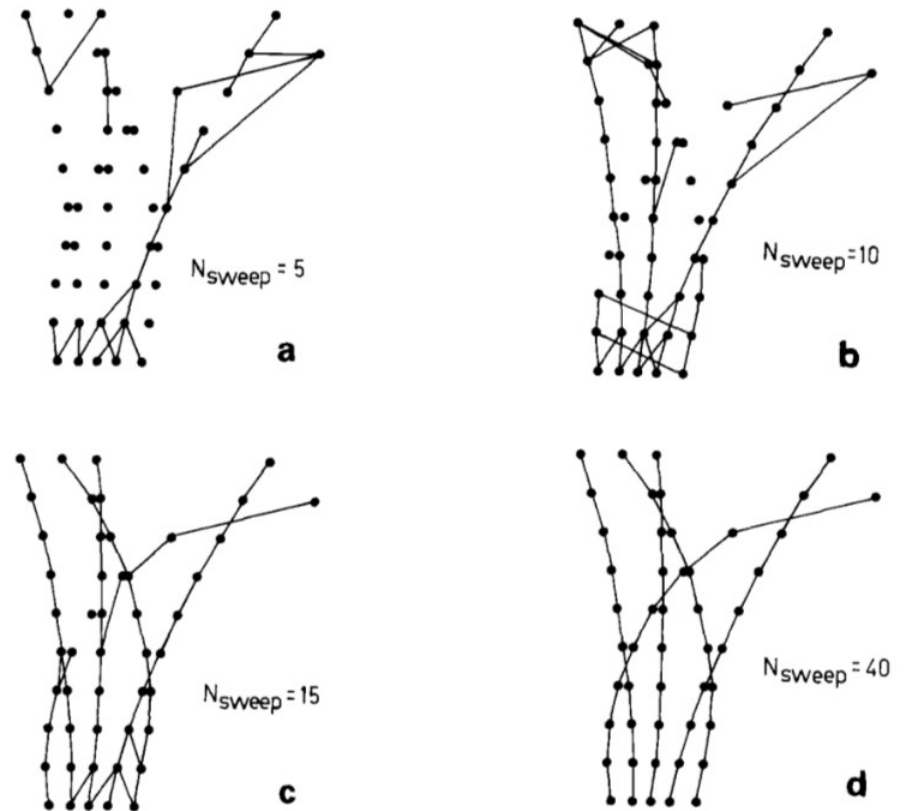
Charged Particle Tracking using Adiabatic Quantum Annealing through doublet classification



Hopfield Network

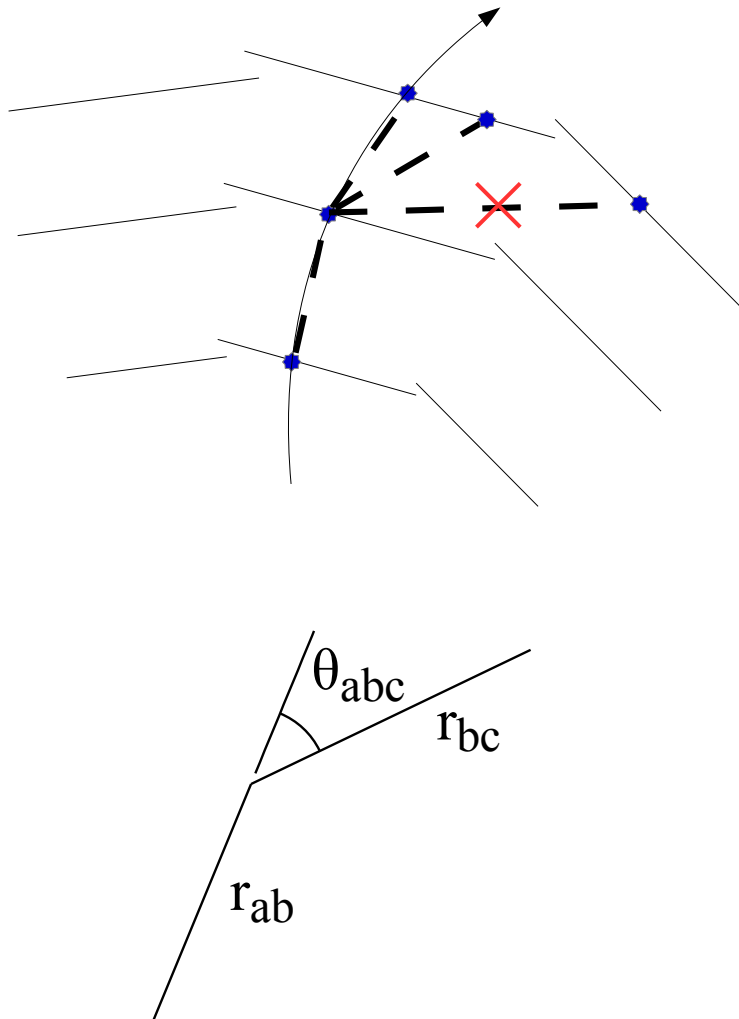
- Developed by John Hopfield in 1982
- fully-connected, single-layer NN;
complete graph
- vertices: n binary units, $\{s_n\} \in \{0, 1\}^n$
- edges: symmetric weight matrix,
 $w \in \mathbb{R}^n \times \mathbb{R}^n$
- energy associated with each network configuration (assignment of units):

$$E = -\frac{1}{2} \sum_{i,j} w_{ij} s_i s_j \quad \text{QUBO!}$$



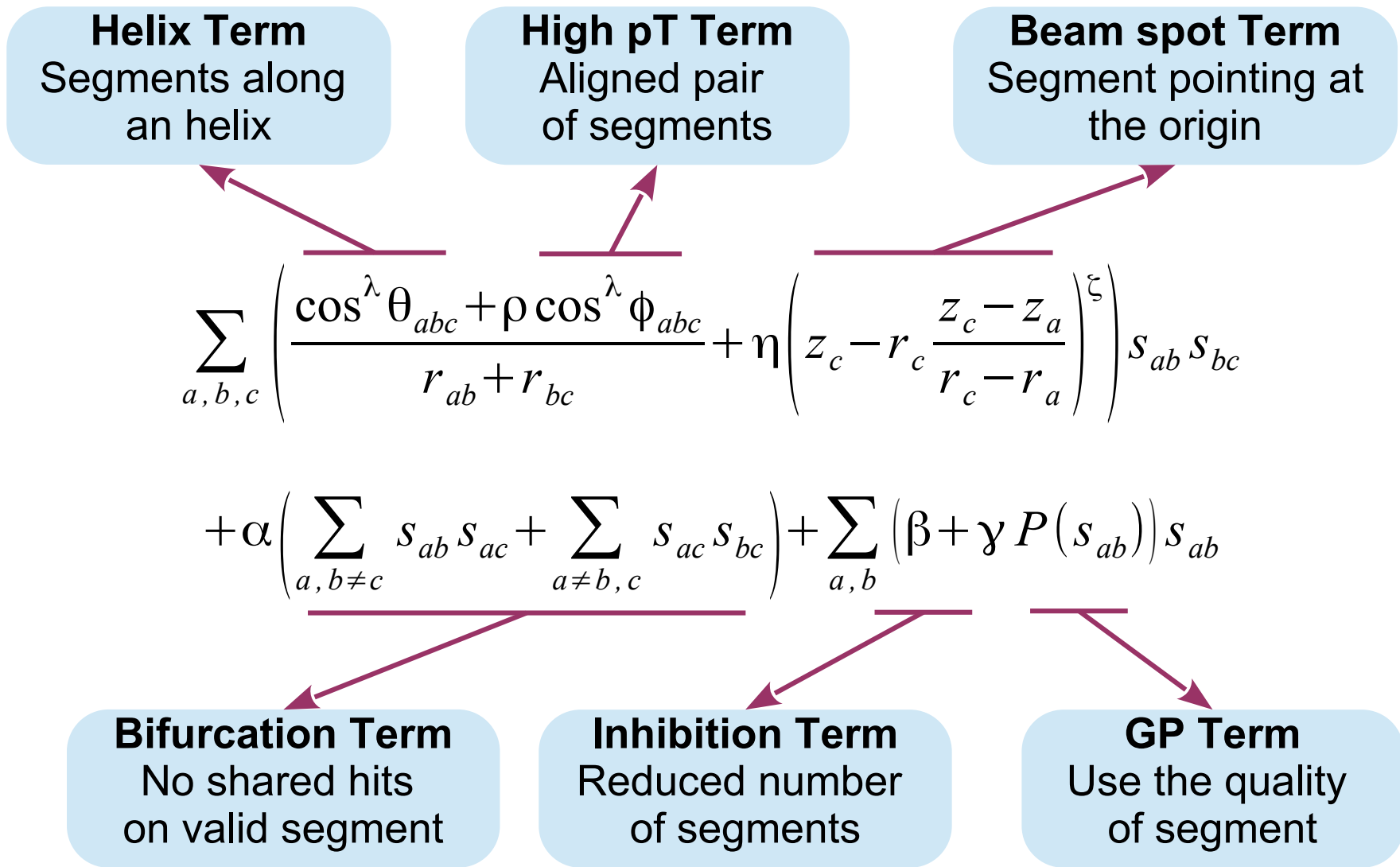
[Peterson, 1989]

Framing the Problem



- **Segment** \equiv pair of hits on consecutive layers of the detector
- Assign a **boolean** to each segment representing whether the segment is within a track or not
- Limits the number of hits/segments
 - No double hit per layer
 - Separating the hits in **16 sector in ϕ**
 - pre-filtering the segments on $\Delta\phi$ and Δz to reduce the number of spurious bad segments
- Segment opening in r-phi-z plane in which helical segments are aligned
- Azimuthal angle in cartesian coordinate in which high p_T tracks segments are straight

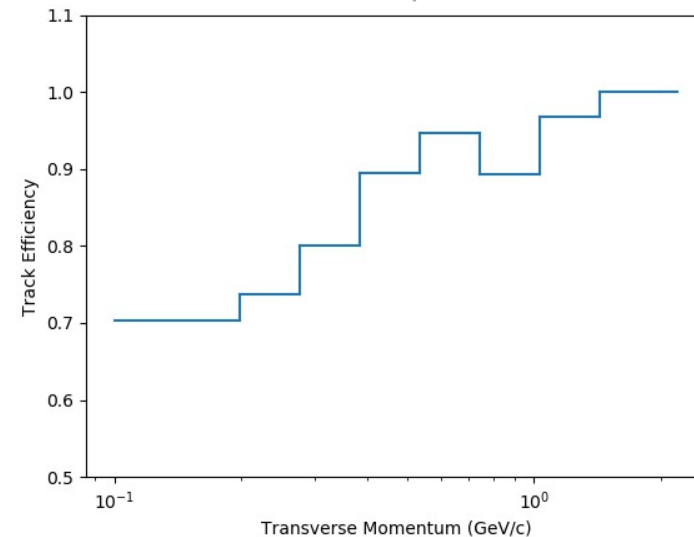
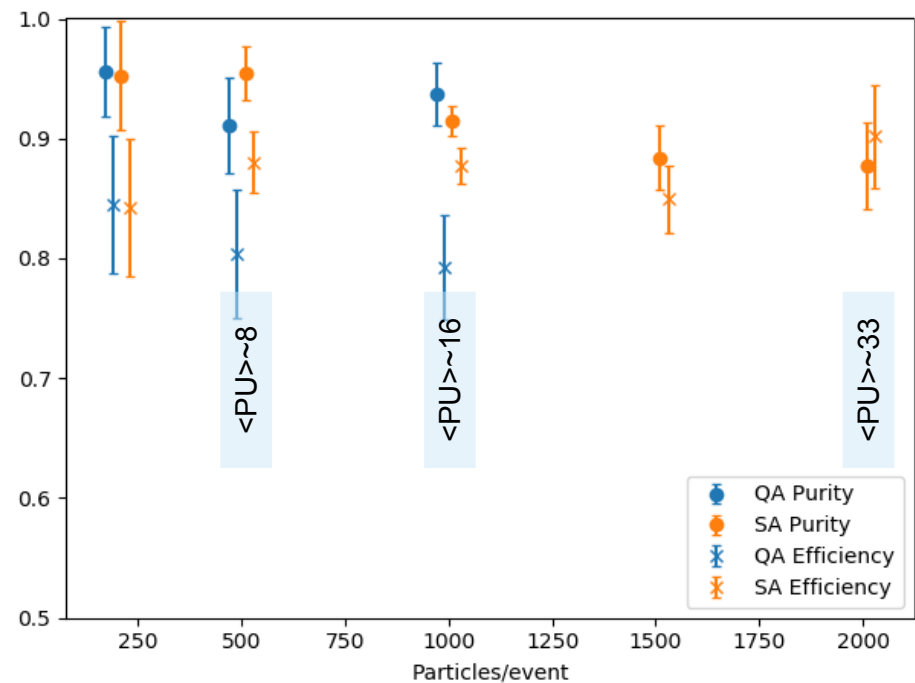
Segment QUBO



Performance

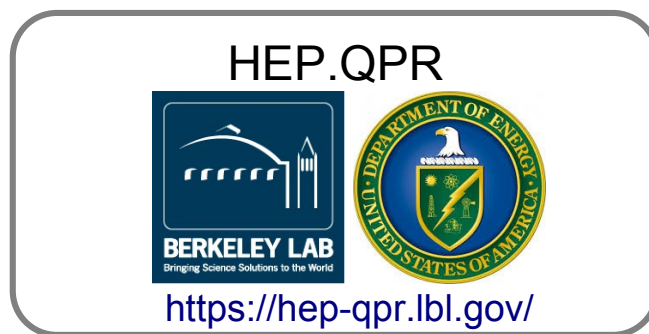


- Simulated annealing and Quantum annealing are in perfect agreement at 200 tracks
- Simulated annealing solves the exact problem at all multiplicity
- Limitation on number of qubits prevents from solving events beyond 200 tracks on Dwave ; solving a contrive problem
- Purity and Efficiency are measured with respect to true tracks with at least three hits
- Promising tracking efficiency for the algorithm up to 2000 tracks per event



Charged Particle Tracking using Adiabatic Quantum Annealing through **triplet** classification

<https://arxiv.org/abs/1902.08324>



- “regard a track with n hits as a set of $n-1$ consecutive lines [doublets] with a smooth shape and without bifurcation”.
- Algorithm
 - Generate the set of potential doublets (apply cleaning cuts)
 - Binary classification task to determine which doublets should be kept in the track candidates, using Hopfield Networks

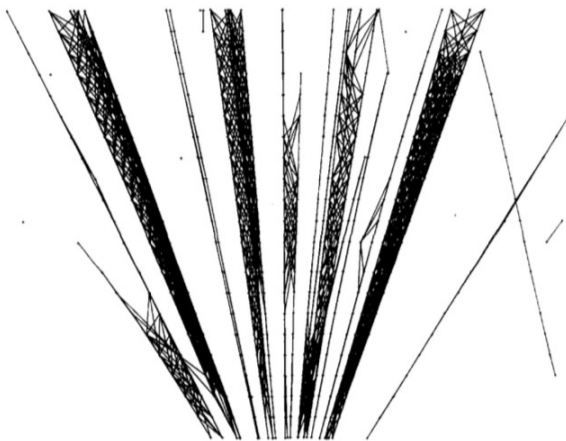


Fig. 2. Display of all generated lines for a real $Z^0 \rightarrow$ hadrons (RZ projection).

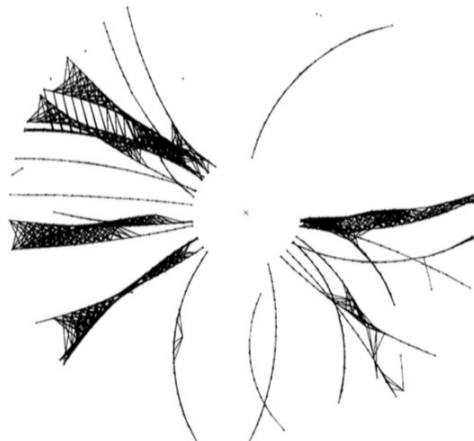


Fig. 1. Display of all generated lines for a real $Z^0 \rightarrow$ hadrons (XY projection).

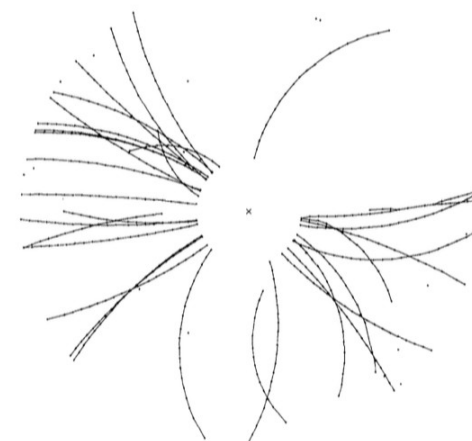


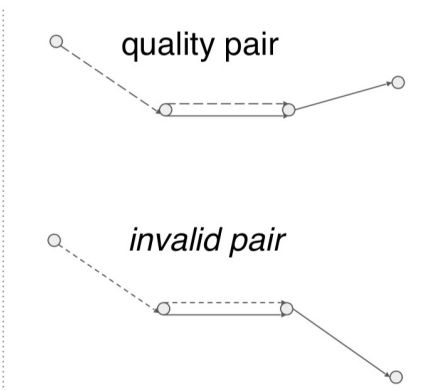
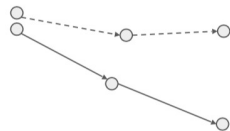
Fig. 3. Display of the activated lines after convergence for a real $Z^0 \rightarrow$ hadrons (XY projection).

<https://www.sciencedirect.com/science/article/pii/S016890021900488P>

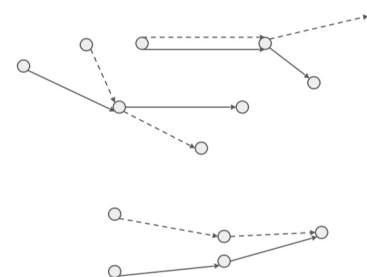
Triplet QUBO



unrelated triplets



conflicting triplets



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

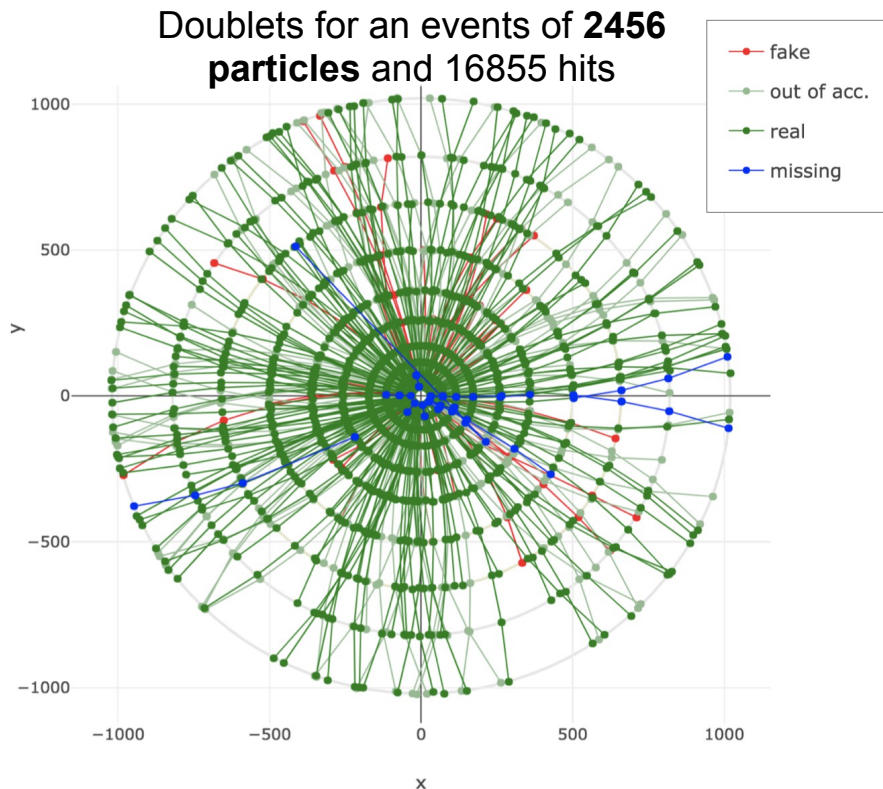
$$b_{ij} = \begin{cases} -S(T_i, T_j), & \text{if } (T_i, T_j) \text{ form a quadruplet,} \\ \zeta & \text{if } (T_i, T_j) \text{ are in conflict,} \\ 0 & \text{otherwise.} \end{cases}$$

$$S(T_i, T_j) = z_1 \frac{z_2 \left(1 - |\delta(q/p_{T_i}, q/p_{T_j})|\right)^{z_3}}{(1 + H_i + H_j)^{z_5}} + \frac{(1 - z_2) \left(1 - \max(\delta\theta_i, \delta\theta_j)\right)^{z_4}}{(1 + H_i + H_j)^{z_5}}$$

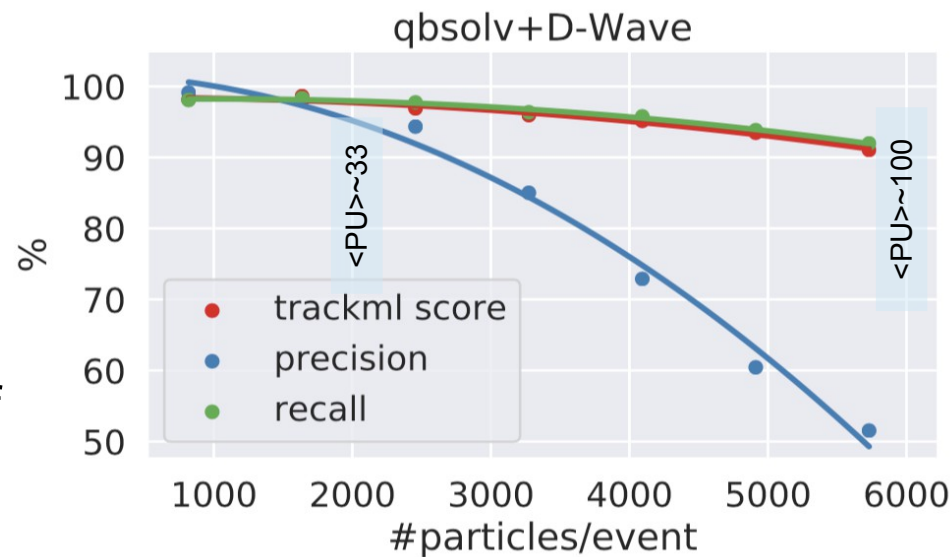
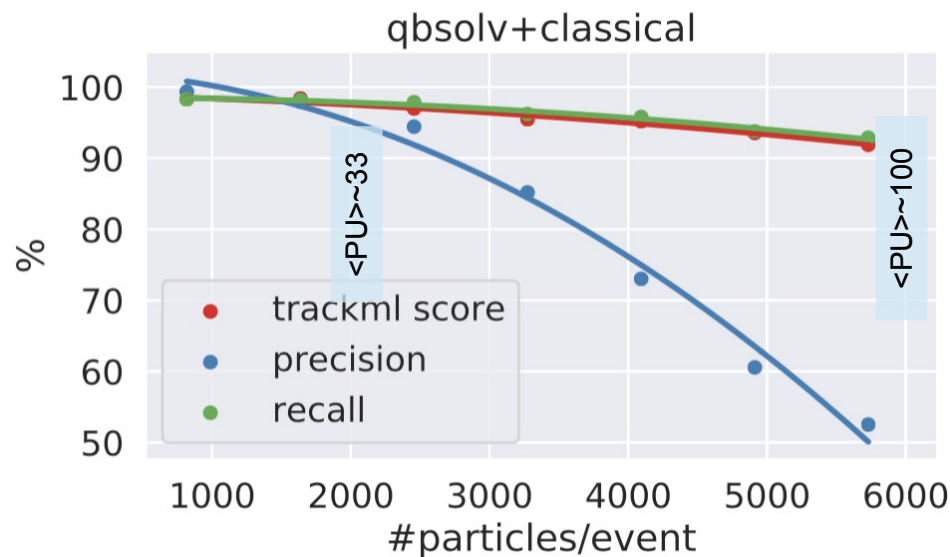
$$S(T_i, T_j) = \frac{1 - \frac{1}{2} \left(|\delta(q/p_{T_i}, q/p_{T_j})| + \max(\delta\theta_i, \delta\theta_j) \right)}{(1 + H_i + H_j)^2}$$

parameters tuned to favour high pt tracks

Performance

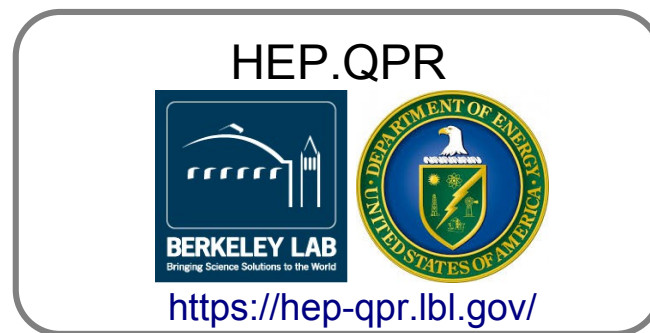


- Full problem resolution using qbsolve : classical simulator, or dwave compared.
- Hits from barrel only
- TrackML score conserved over range of event density.
- Improved performance (not shown) with more information (IP constraint, ...)

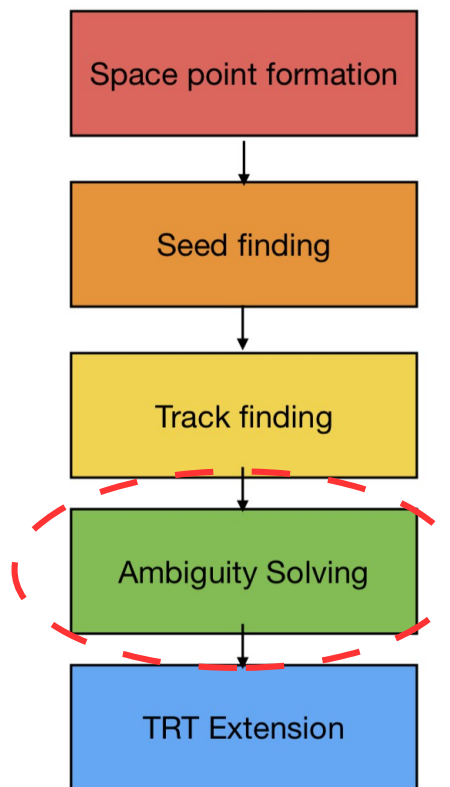
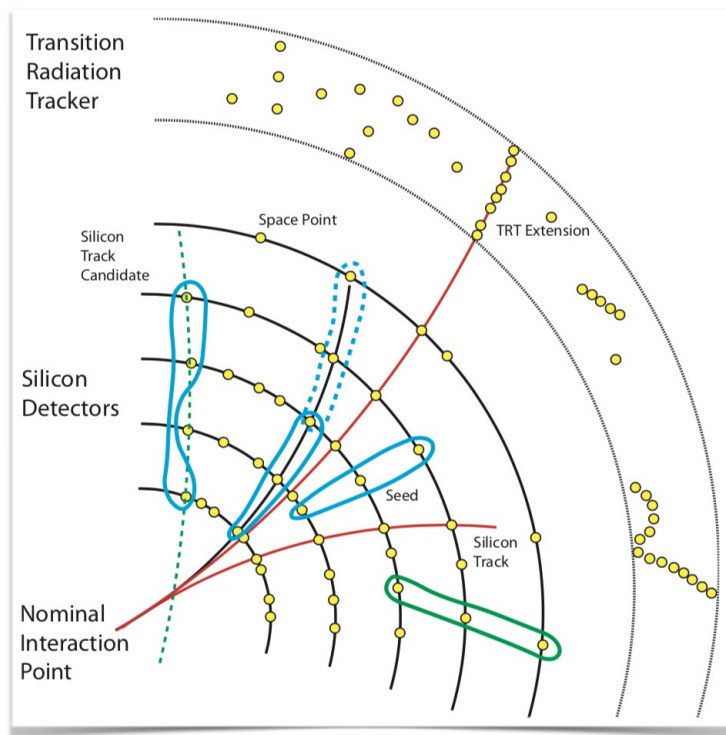


<https://arxiv.org/abs/1902.08324>

Track Ambiguity Resolution using Adiabatic Quantum Annealing



Resolving Close-by Tracks



- Loosen constraints during pattern recognition to increase efficiency.
 - Reduce fake rate by resolving ambiguities.
 - Particularly relevant in dense environment like the core of a jet.
- Early results on resolving ambiguities from a solution of the trackML challenge : 15% purity and 70% efficiency

Conclusion



- The dataset of the TrackML Challenge is a valuable resource for the community.
- It is early for quantum computing but the exploration has begun, to prepare and think differently on algorithms.
- Promising application of Quantum annealer to charged particle tracking tasks.



QMLQCF Ackn.



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Part of this work was conducted at "iBanks", the AI GPU cluster at Caltech. We acknowledge NVIDIA, SuperMicro and the Kavli Foundation for their support of "iBanks".



Extra Material



The D-Wave Company



D:wave
The Quantum Computing Company™

COMPANY ▾

TECHNOLOGY ▾

COMPUTING ▾

RESOURCES ▾

NEWS ▾

Welcome to the Future

Quantum Computing for the Real World Today

<https://www.dwavesys.com/>

- 1999** Founded
- 2011** D-Wave One : 128 qubits
- 2013** D-Wave Two : 512 qubits
- 2015** D-Wave 2X : 1000 qubits
- 2017** D-Wave 2000Q : 2000 qubits
- 2019?** 5000 qubits ?

D-Wave Hamiltonian And Chimera Graph

D-Wave Hamiltonian



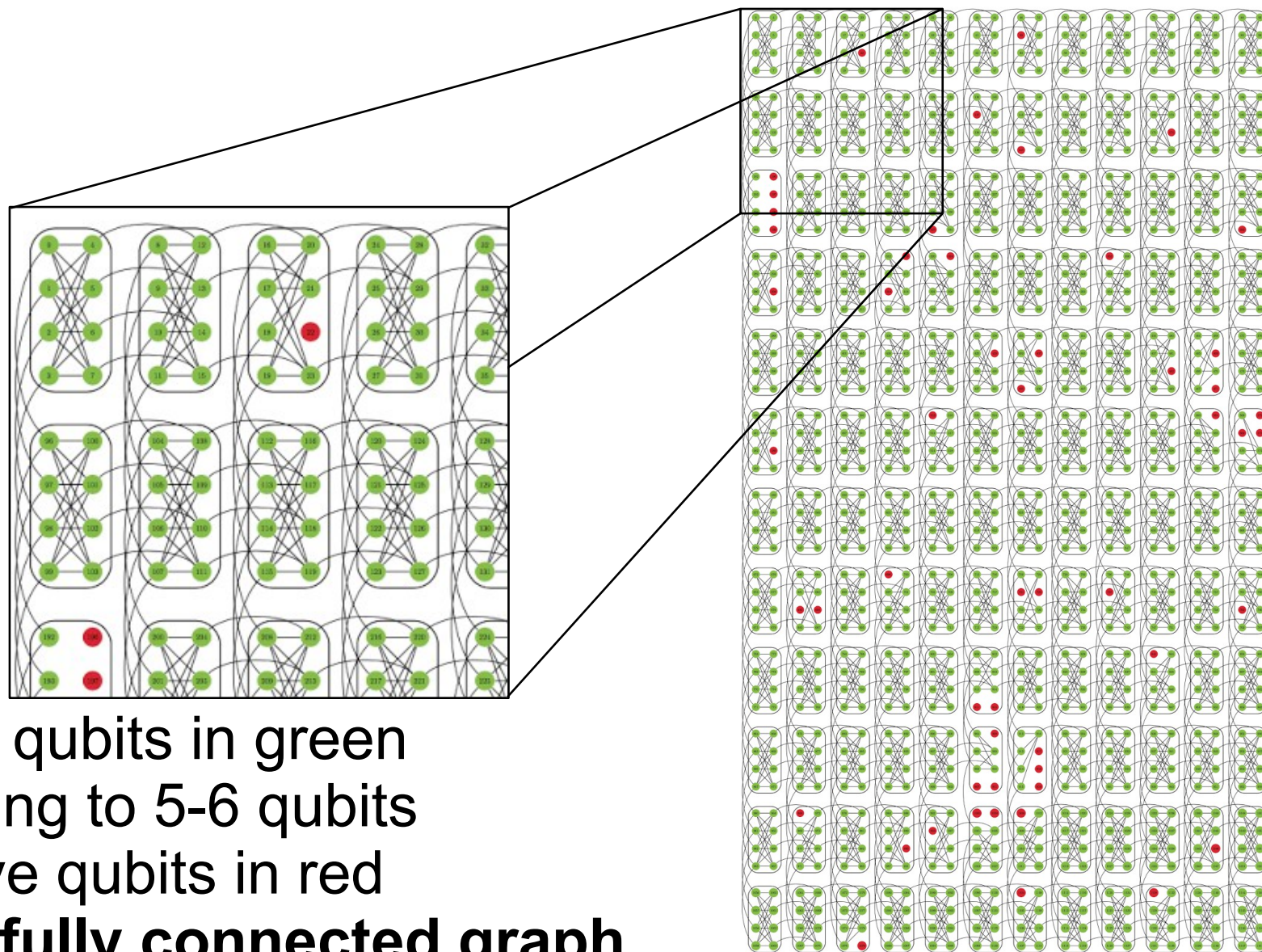
$$H_{\text{Ising}} = \sum_i h_i \sigma_i^z + \sum_{ij} J_{ij} \sigma_i^z \sigma_j^z$$

Runs over adjacent quBits

External magnetic field

Interactions

D-Wave qubit Adjacency



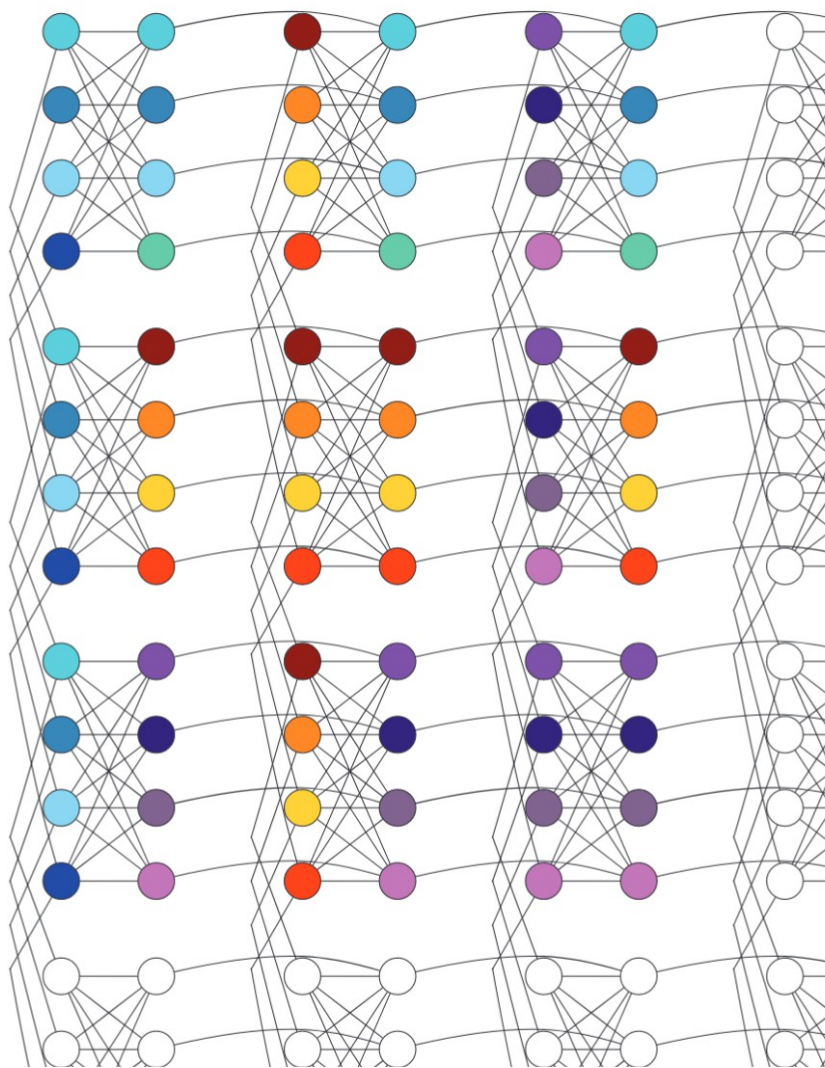
Active qubits in green
Coupling to 5-6 qubits
Inactive qubits in red
Not a fully connected graph



Model Embedding



Full Ising Model



- Create chains of spins through the chimera graph
 - Split local fields across all qubits in the chain
 - Tightly couple ($J_F=6$)
 - Non-unique embedding. Heuristic approach.
 - Suppressing spin flip within chain as error correction.
 - Use majority vote
- Approximately full Ising Model with $\sim <40$ spins

<https://arxiv.org/abs/1210.8395>

Ising Hamiltonian



$$H_{\text{Ising}} = \sum_i h_i \sigma_i^z + \sum_{ij} J_{ij} \sigma_i^z \sigma_j^z$$

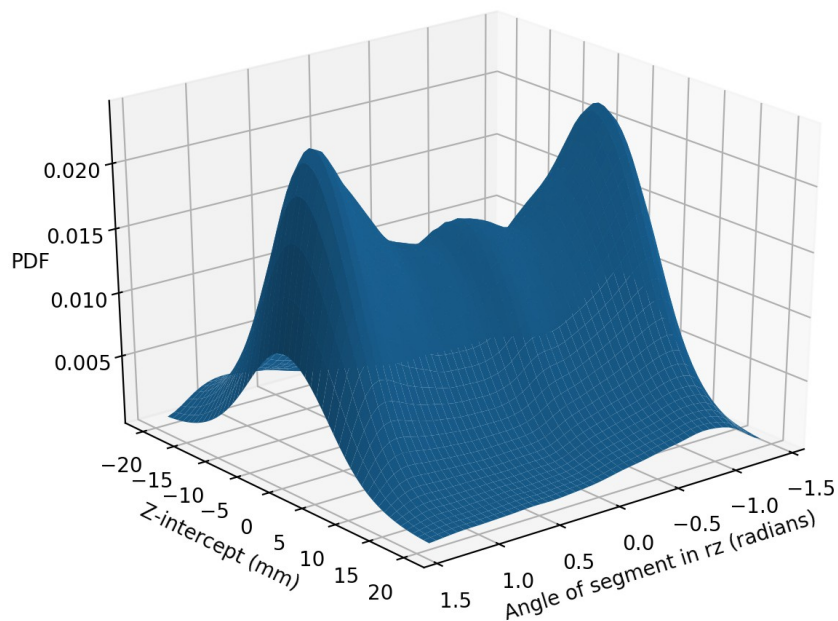
Runs over **all** quBit pairs

External magnetic field

Interactions

Resolving Sub-Group

- Full all-to-all QUBO problem cannot fit on dWave. Aim at identifying **sub-groups of segments that can fit on the hardware**



- Train a gaussian kernel density estimator on true single segment
- Aiming at reducing the number of false segments, retaining
- Force segment off based on $\cos\theta_{abc}$
 - $\cos\theta_{abc} = 0$ if $\theta_{abc} > \theta_0$
- 5 best neighbors
- Solvable in polynomial time

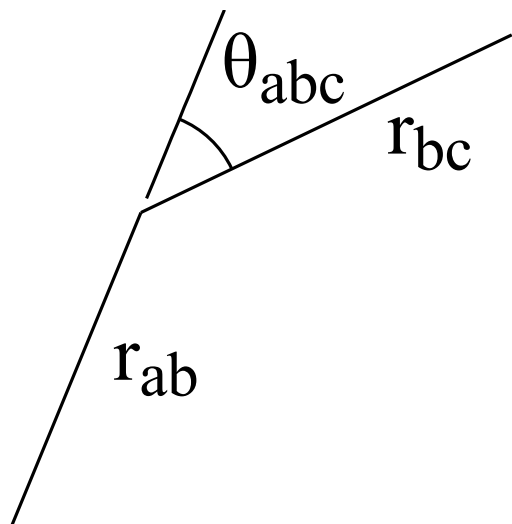


Problem Parameters Optimization

- Parameters of the hamiltonian are tuned using bayesian optimization, modeling the figure of merit with gaussian processes.
- Accuracy (# of properly labeled / # of segments) use as f.o.m
- Global inhibition model : $\alpha=3.E^{-3}$, $\beta=2.63E^{-8}$, $\lambda=7$
- Threshold model : $\alpha=5.E^{-3}$, $\beta=1.E^{-6}$, $\lambda=7$

Edge Affinity

- Helical bias: tracks are straight in cylindrical coordinates
- Momentum bias: high-PT tracks are straight in rectangular coordinates
- Short-edge bias: long tracks of short edge segments



Helical bias
(cylindrical angle)

Momentum bias
(rectangular angle)

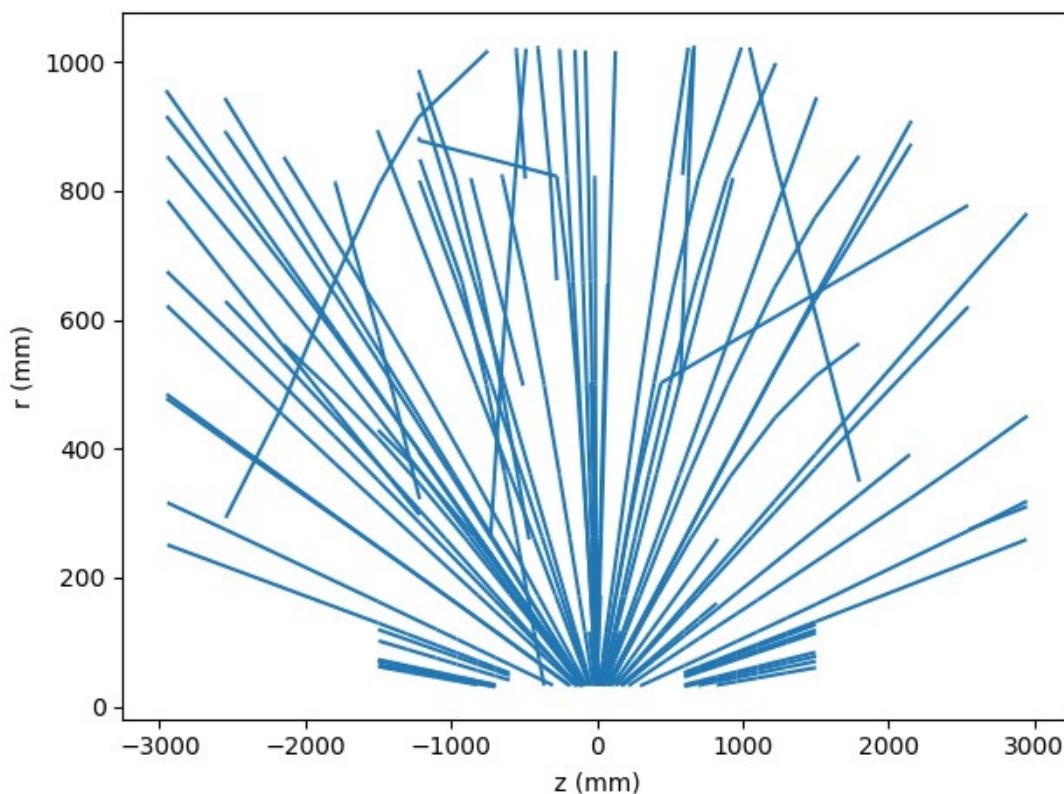
$$\sum_{a,b,c} \frac{\cos^\lambda \theta_{ab} + \rho \cos^\lambda \phi_{ab}}{r_{ab} + r_{bc}} S_{ab} S_{bc}$$

Short-edge bias

Ising variables (1 or 0)

Cross-Term Penalties

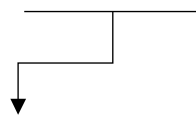
- Beam spot penalty: penalize tracks that originate further from the origin



Z-intercept penalty



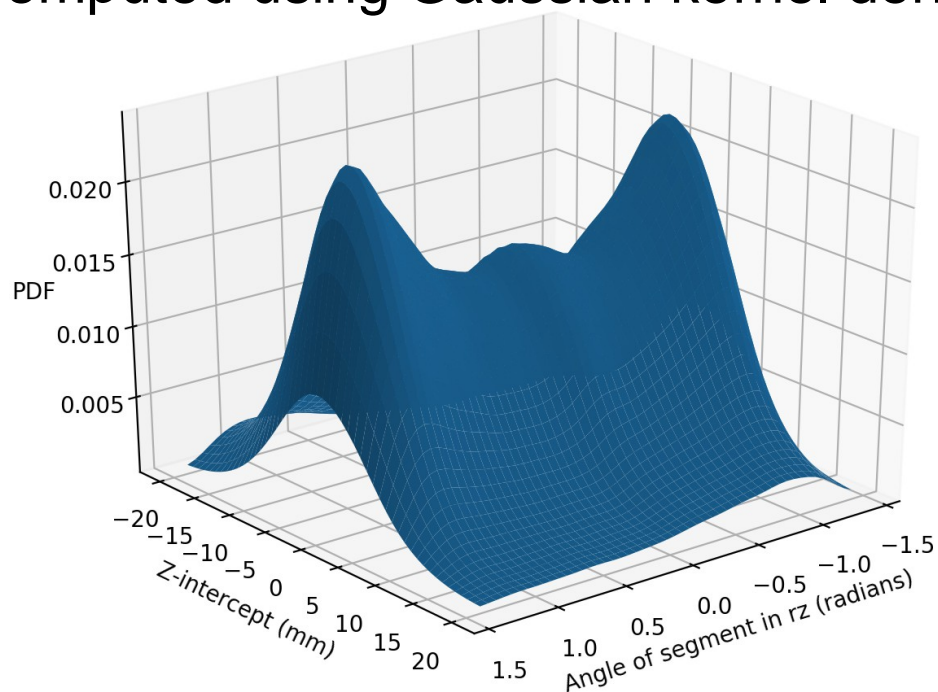
$$\sum_{a,b,c} \eta \left(z_c - \frac{z_c - z_a}{r_c - r_a} r_c \right)^\zeta s_{ab} s_{bc}$$



Ising variables (1 or 0)

Single-Edge Bias

- Global inhibition: limits total number of edges turned on
- Prior probability: Bayesian prior based on edge position in rz -plane
 - Computed using Gaussian kernel density estimation

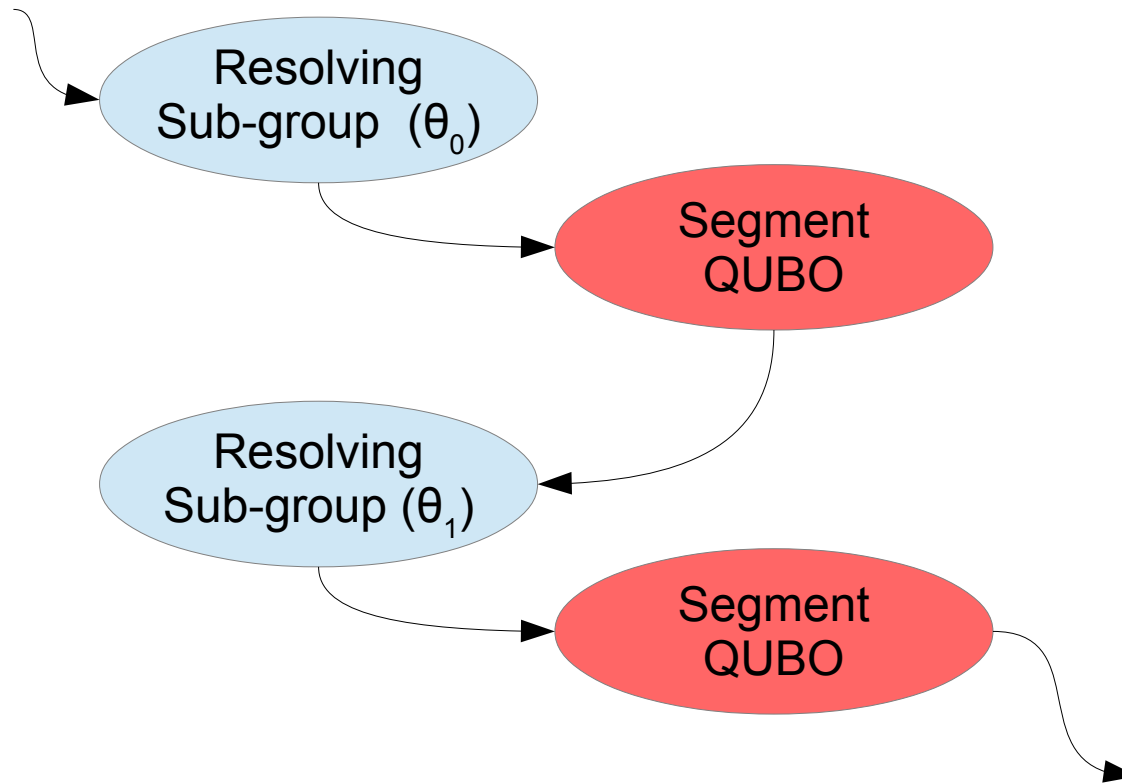


Global inhibition Bayesian prior

$$\sum_{a,b} [\beta + \gamma P(s_{ab})] s_{ab}$$











Ising variable (1 or 0)

QA-Tracking Workflow

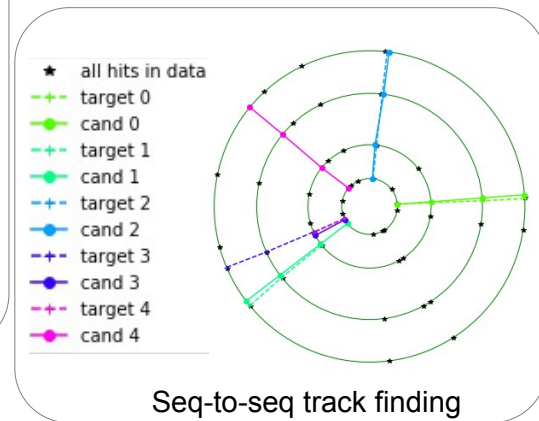
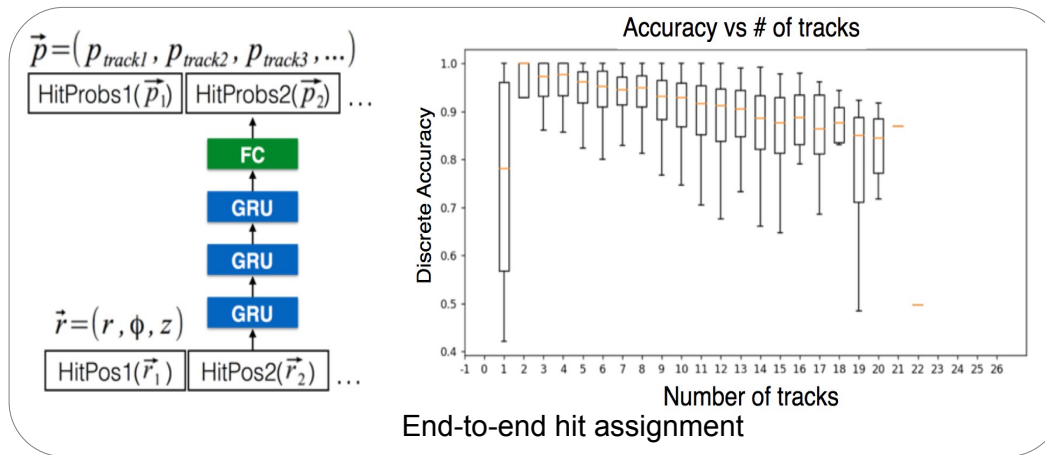
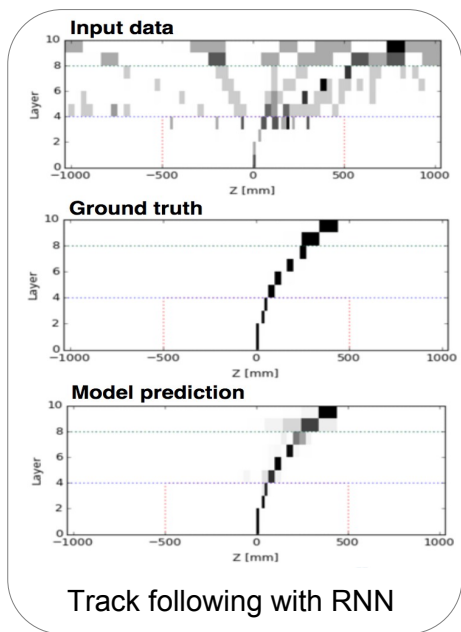


References

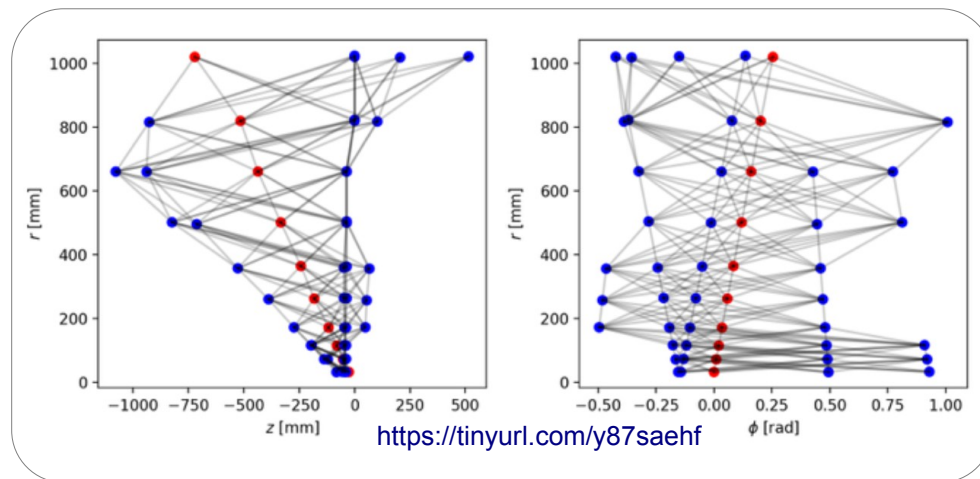
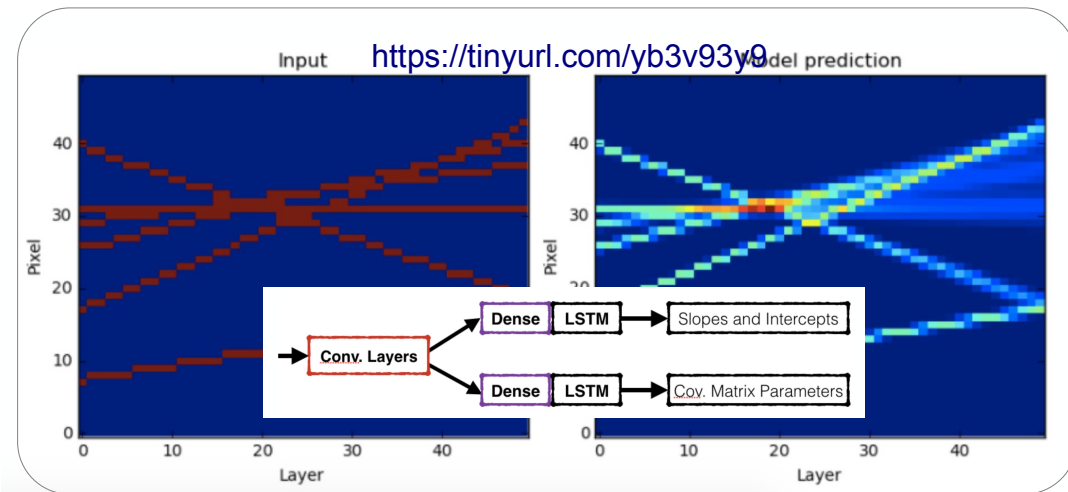


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HEP.TrkX Approaches



<https://heptrkx.github.io/>



Hits in the barrel region, track with a least 5 hits,
and 1 GeV or pT, no double hit per layer.

potential
doublets

