

# Quantum Algorithms for Jet Clustering

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(arXiv:1908.08949)

DOE CSGF Program Review

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# Outline

Introduction

Thrust

Classical Algorithms

Quantum Annealing

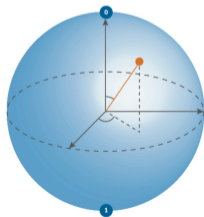
Quantum Search

Towards Jet Algorithms for the LHC

Conclusion and Future Directions

# Quantum Computing

- ▶ Definition: Using quantum systems to perform computations.
- ▶ Challenge: How do we identify problems where we get a speedup over classical computers?
- ▶ Main approaches:
  - ▶ Analog
    - ▶ Quantum simulation
    - ▶ Quantum annealing
  - ▶ Digital: define operations in terms of logic gates on qubits.



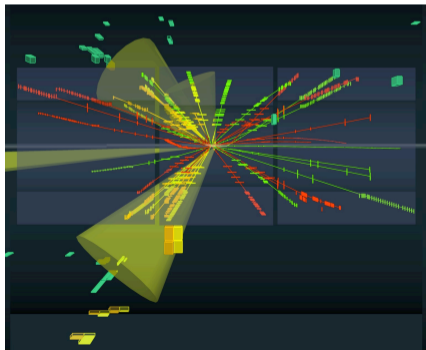
# Why Quantum Computing for Particle Physics?

DOE QuantISED initiative:

- A. Cosmos and Qubits: applications of quantum error correction, scrambling, computational complexity to black holes and holography
- B. Foundational QIS-HEP Theory and Simulations: applications of quantum simulation; applications of quantum information and entanglement to field theory techniques
- C. Quantum Computing for HEP: applications of quantum machine learning and data analysis tools
- D. QIS-based Quantum Sensors: applications of quantum sensors to experiments
- E. Research Technology for QIST: apply accelerator technology to qubit technology

## Goals of This Work

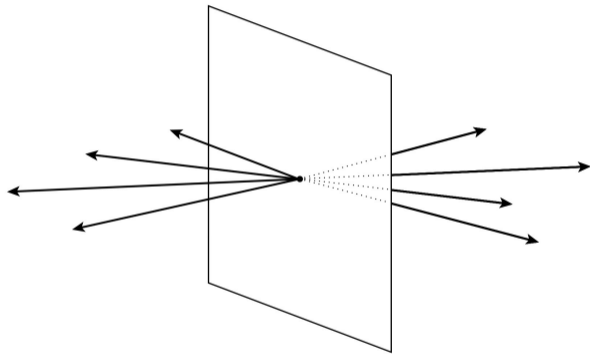
- ▶ Jets are collections of collimated, energetic hadrons formed in high-energy collisions.
- ▶ Identifying jets can be computationally intensive. Can quantum computers help?
- ▶ What would a realistic application of quantum search look like?
- ▶ Goal #1: investigate quantum algorithms. ✓
- ▶ Goal #2: improve classical algorithms. ✓



## Case Study: Thrust

(Brandt, Peyrou, Sosnowski, Wroblewski '64; Farhi '77)

- ▶ For  $e^+e^-$  collision, find optimal separating plane to partition into two jets.



- ▶ Why thrust?
  - ▶ Thrust is a global optimization problem!
  - ▶ anti- $k_t$  (default LHC algorithm) is  $O(N \log N)$ , vs  $O(N^3)$  for thrust, but it is a local heuristic.

# Quantum Computing

We will apply two key quantum computing paradigms.

- ▶ Quantum annealing: used in commercial hardware like D-Wave.
- ▶ Grover search: workhorse algorithm in quantum gate model.
  - ▶ Grover search finds desired item in size  $N$  database using only  $O(\sqrt{N})$  time.
  - ▶ This is not realistic! Loading data that lives in classical memory takes at least  $O(N)$  time.
  - ▶ To obtain speedup, size of search space needs to scale like  $O(N^\alpha)$ , with  $\alpha$  offsetting data loading cost.

# Results

Implementation	Time Usage	Qubit Usage
Classical (Yamamoto '84)	$O(N^3)$	—
Classical: Sort	$O(N^2 \log N)$	—
Classical: Parallel Sort	$O(N \log N)$	—
Quantum Annealing	Gap Dependent	$O(N)$
Quantum Search: Sequential	$O(N^2)$	$O(\log N)$
Quantum Search: Parallel	$O(N \log N)$	$O(N \log N)$

- ▶ We improve the classical thrust algorithm using a trick from (Salam, Soyez '07)
- ▶ We consider two quantum data loading models, **sequential** (fewer qubits but more time per query) and **parallel** (more qubits and less time per query).



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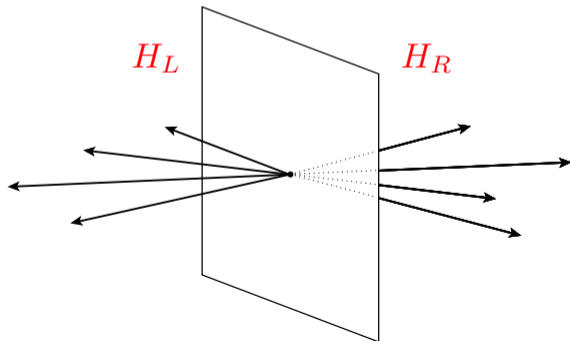
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## Definition of Thrust: Partitioning Problem

- ▶ Given  $N$  three-momenta  $\{\vec{p}_i\}$  in CM frame, find partition  $H_L \cup H_R$  maximizing

$$T(H_L) = \frac{2 \left| \sum_{i \in H_L} \vec{p}_i \right|}{\sum_{i=1}^N |\vec{p}_i|} = \frac{2 \left| \sum_{i \in H_R} \vec{p}_i \right|}{\sum_{i=1}^N |\vec{p}_i|}.$$

- ▶ Brute-force classical strategy:  $O(N 2^N)$ . Motivates quantum annealing algorithm.

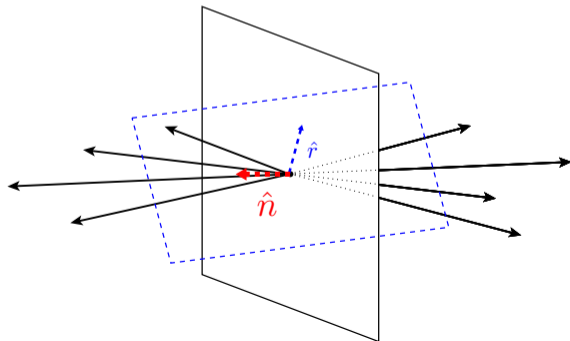


## Definition of Thrust: Axis-Finding Problem

- ▶ For  $\hat{n}$  a unit norm vector, find  $\hat{n}_{opt}$  maximizing

$$T(\hat{n}) = \frac{\sum_{i=1}^N |\hat{n} \cdot \vec{p}_i|}{\sum_{i=1}^N |\vec{p}_i|} = \frac{2 \sum_{i=1}^N \Theta(\hat{n} \cdot \vec{p}_i) (\hat{n} \cdot \vec{p}_i)}{\sum_{i=1}^N |\vec{p}_i|}.$$

- ▶ Partitions particles into  $\hat{n}_{opt} \cdot \vec{p}_i > 0$  and  $\hat{n}_{opt} \cdot \vec{p}_i < 0$ .
- ▶ Motivates classical  $O(N^3)$  algorithm and Grover search algorithm.



## Definition of Thrust: Duality

(Thaler '15)

- ▶ These two definitions of thrust are in fact equivalent.
- ▶ Can be shown by directly optimizing, over partition  $H$  with  $\vec{P} = \sum_{i \in H} \vec{p}_i$ , the objective function

$$O(\vec{P}, \hat{n}) = \hat{n} \cdot \vec{P} + \lambda(\hat{n}^2 - 1).$$

- ▶ Optimize over  $\hat{n}$  first:

$$\hat{n}_{opt} = \frac{\vec{P}}{|\vec{P}|} \Rightarrow O(\vec{P}, \hat{n}_{opt}) = |\vec{P}|.$$

- ▶ Optimize over  $\vec{P}$  first:

$$\vec{P}_{opt} = \sum_{i=1}^N \Theta(\hat{n} \cdot \vec{p}_i) \vec{p}_i \Rightarrow O(\vec{P}_{opt}, \hat{n}) = \sum_{i=1}^N \Theta(\hat{n} \cdot \vec{p}_i) (\hat{n} \cdot \vec{p}_i).$$

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# Classical Thrust Algorithm

(Yamamoto '84) Finds axis defining the separating plane.

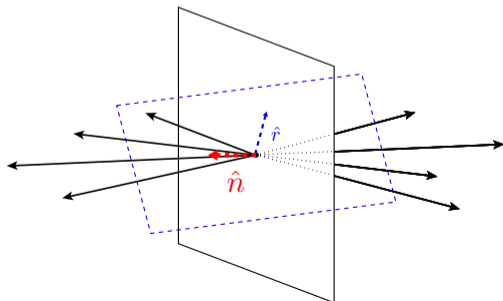
1. For each pair  $\vec{p}_i, \vec{p}_j$ :

▶ Compute reference vector  $\hat{r}_{ij} = \frac{\vec{p}_i \times \vec{p}_j}{|\vec{p}_i \times \vec{p}_j|}$ .

▶ For each  $\vec{p}_k$ , assign to partition  $H_{ij}$  if  $\hat{r}_{ij} \cdot \vec{p}_k > 0$ , and ignore otherwise.

2. Return  $\max T(H_{ij})$ .

▶ Requires time  $O(N^3) = \underbrace{O(N^2)}_{\text{Number of } H_{ij}} \times \underbrace{O(N)}_{\text{Sum the } \vec{p}_k \text{ for each } H_{ij}}$ .



# Classical Thrust Algorithm: Improvements via Sort

(Using Salam, Soyez '07)

- ▶ Traverse particles in special order to avoid redundant  $O(N)$  cost for each  $H_{ij}$ .

1. For each  $\vec{p}_i$ ,

- ▶ Sort  $\vec{p}_j$  by azimuth b/w  $\vec{p}_j$  and  $\vec{p}_i$ , obtaining  $\{\vec{p}_{j_1}, \vec{p}_{j_2}, \dots\}$ .
- ▶ For  $\vec{p}_{j_1}$ , loop over all  $\vec{p}_k$ , assigning to partition  $H_{ij_1}$  if  $\hat{r}_{ij_1} \cdot \vec{p}_k > 0$ .
- ▶ For  $\vec{p}_{j_n}$ ,  $n \geq 2$ , obtain  $T(H_{ij_n})$  from  $T(H_{ij_{n-1}})$  using the fact that only one particle ever enters and leaves  $H_{ij_{n-1}}$ .

- ▶ Requires time

$$O(N^2 \log N) = \underbrace{O(N)}_{\text{Number of } \vec{p}_i} \times \left[ \underbrace{O(N \log N)}_{\text{Sort}} + \underbrace{O(N)}_{\text{Compute } T(H_{ij_1})} + \underbrace{O(N)}_{\text{Compute } T(H_{ij_n}), n \geq 2} \right].$$

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# Quantum Annealing



- ▶ Existing hardware like D-Wave are quantum annealers.
- ▶ Hardware consists of spins on a lattice; challenge is controlling connectivity between spins.
- ▶ Solution to optimization problem is encoded in the ground state of an Ising-like Hamiltonian.

## Quantum Annealing: Thrust (1/2)

- ▶ Equivalently, for  $x_i \in \{0, 1\}$ , we want to minimize the QUBO objective function

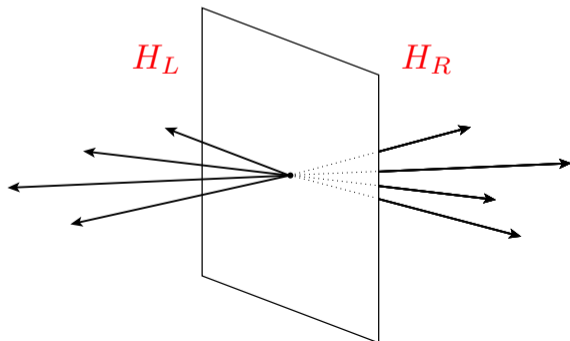
$$O(\{x_i\}) = \sum_{i,j=1}^N Q_{ij} x_i x_j.$$

- ▶ Remarkably, thrust can be written in QUBO form!
- ▶ Recall definition of thrust as a partitioning problem.

## Quantum Annealing: Thrust (2/2)

- ▶ Partition  $H$  has total three-momentum  $\vec{P}(\{x_i\}) = \sum_{i=1}^N \vec{p}_i x_i$ , where  $x_i = 1$  for  $i \in H$  and  $x_i = 0$  otherwise. Then  $T(H) \propto |\vec{P}|$ .
- ▶ This is monotonic  $\Rightarrow$  square to obtain QUBO function:

$$O(\vec{x}_i) = -|\vec{P}|^2 = -\sum_{i,j=1}^N (\vec{p}_i \cdot \vec{p}_j) x_i x_j.$$



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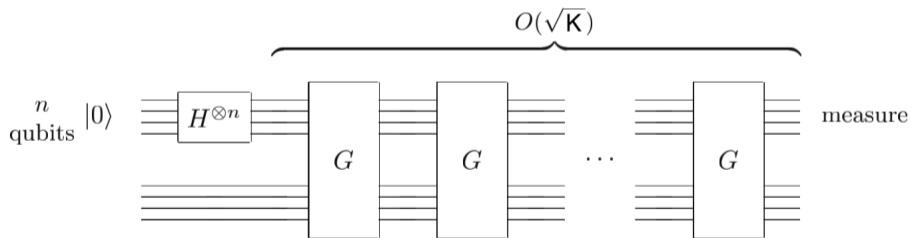
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## Quantum Search (1/2)

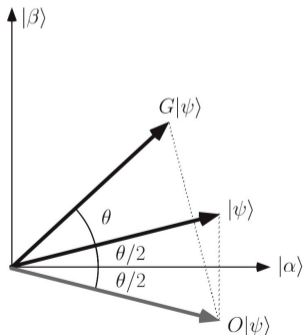
- ▶ Search is the backbone of optimization algorithms.
- ▶ Grover search performs search in the quantum gate model.
  - ▶ In this model, algorithm = prepare initial state, apply unitary gates, measure to obtain desired final state.
  - ▶ Grover uses  $\sqrt{K}$  gates to search  $K$  items.



## Quantum Search (2/2)

Grover search (Grover '96; Boyer, Brassard, Hoyer, Tapp '98):

1. Prepare initial state  $|\psi_0\rangle = \frac{1}{\sqrt{K}} \sum_{i=1}^K |i\rangle$ .
2. Repeat  $O(\sqrt{K})$  times:
  - ▶ Reflect about marked states.
  - ▶ Reflect about initial state  $|\psi_0\rangle$ .



# Quantum Search: Maximum Finding

- ▶ Maximum finding algorithm (Durr, Hoyer '96): Return max of array using Grover search.
  - ▶ Track best max so far; marked states are those larger than current max.
- ▶ We will apply this algorithm to thrust search space of size  $K = O(N^2) \Rightarrow$  outer Grover loop scales like  $O(N)$ .

# Quantum Search: Thrust Algorithm (1/2)

Schematically,

1. Pick random indices  $m, n$  corresponding to current max.

2. Repeat  $O(N)$  times:

▶ Prepare initial state  $|\psi_0\rangle = \frac{1}{2^N} \sum_{i,j=1}^{2^N} |i\rangle |j\rangle |0\rangle |0\rangle |\hat{0}\rangle |0\rangle |0\rangle$ .

▶ Reflect about marked states:

2.1 Compute  $T_{ij}$ , mapping  $|i\rangle |j\rangle |0\rangle |0\rangle |\hat{0}\rangle |0\rangle |0\rangle \mapsto |i\rangle |j\rangle |0\rangle |0\rangle |\hat{0}\rangle |0\rangle |T_{ij}\rangle$ .

2.2 Apply phase to states with  $T_{ij} > T_{mn}$ , mapping

$|i\rangle |j\rangle |0\rangle |0\rangle |\hat{0}\rangle |0\rangle |T_{ij}\rangle \mapsto (-1)^{\Theta(T_{ij}-T_{mn})} |i\rangle |j\rangle |0\rangle |0\rangle |\hat{0}\rangle |0\rangle |T_{ij}\rangle$ .

2.3 Uncompute intermediate registers.

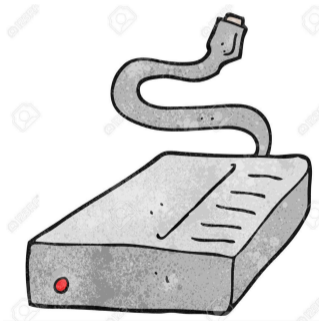
▶ Reflect about initial state.

▶ Measure  $i, j$  register. If  $T_{ij} > T_{mn}$  update  $m, n$ .



## Quantum Search: Data Loading (1/2)

- ▶ Grover algorithm assumed ability to prepare superposition over all items.
- ▶ Collider data lives on classical hard drive, not in quantum superposition!



## Quantum Search: Data Loading (2/2)

- ▶ Data loading needs to happen in the step that computes  $T_{ij}$ .
- ▶ Define two abstract unitary operations:
  - ▶ LOOKUP:  $U_{LOOKUP} |i\rangle |\vec{0}\rangle = |i\rangle |\vec{p}_i\rangle$ .
  - ▶ SUM:  $U_{SUM} |c\rangle |0\rangle = |c\rangle |\sum_{k=1}^N f(\vec{p}_k; c)\rangle$ .
- ▶ Essentially, the data on the hard drive tell us which unitaries to apply.

## Quantum Search: Thrust Algorithm (2/2)

How to compute  $T_{ij}$  using LOOKUP and SUM:

1. Load  $\vec{p}_i, \vec{p}_j$  using LOOKUP:  $|i\rangle |j\rangle |\vec{0}\rangle |\vec{0}\rangle |\hat{0}\rangle |\vec{0}\rangle |0\rangle \mapsto |i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{0}\rangle |\vec{0}\rangle |0\rangle$ .
2. Calculate  $\hat{r}_{ij}$  using  $\vec{p}_i, \vec{p}_j$ :  $|i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{0}\rangle |\vec{0}\rangle |0\rangle \mapsto |i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{0}\rangle |0\rangle$ .
3. Apply SUM with  $f(\vec{p}_k; \hat{r}_{ij}) = \{\vec{p}_k \text{ if } \hat{r}_{ij} \cdot \vec{p}_k > 0; \vec{0} \text{ if } \hat{r}_{ij} \cdot \vec{p}_k < 0\}$  to sum momentum in  $H_{ij}$ :  $|i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{0}\rangle |0\rangle \Rightarrow |i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{P}_{ij}\rangle |0\rangle$ .
4. Divide by normalization factors to obtain thrust:  
 $|i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{P}_{ij}\rangle |0\rangle \mapsto |i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{P}_{ij}\rangle |T_{ij}\rangle$ .
5. Uncompute intermediate registers:  
 $|i\rangle |j\rangle |\vec{p}_i\rangle |\vec{p}_j\rangle |\hat{r}_{ij}\rangle |\vec{P}_{ij}\rangle |T_{ij}\rangle \mapsto |i\rangle |j\rangle |\vec{0}\rangle |\vec{0}\rangle |\hat{0}\rangle |\vec{0}\rangle |T_{ij}\rangle$ .

Runtime:  $O(N(C_{LOOKUP} + C_{SUM}))$ .

## Quantum Search: Sequential Model

- ▶ LOOKUP and SUM require  $O(N)$  time and  $O(\log N)$  qubits.
- ▶ LOOKUP has one register of size  $O(\log N)$  to store  $i, \vec{p}_i$ .
- ▶ LOOKUP scans all  $N$  items in classical database, returning  $\vec{p}_i$ .
- ▶ SUM scans all  $N$  items in classical database to compute  $\sum_{i=1}^N f(\vec{p}_i; c)$ .
- ▶ Thrust algorithm requires  $O(N^2)$  time and  $O(\log N)$  qubits.

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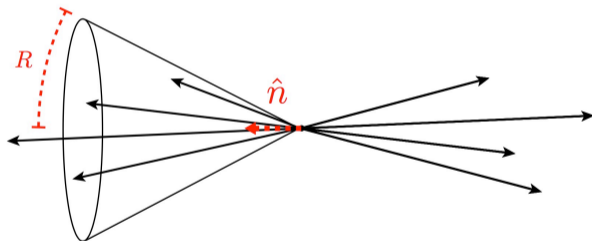
**Towards Jet Algorithms for the LHC**

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# Towards Jet Algorithms for the LHC

How does the thrust case differ from LHC jet algorithms?

- ▶ Different coordinate system from  $e^+e^-$  collisions.
- ▶ Divide event into single jet with opening angle  $R$ , plus unclustered region:



- ▶ Thrust case is  $R = \pi/2$ .

# Jet Function Maximization

(Georgi '14)

- ▶ Find the subset of particles that maximizes the following function ( $\beta = 1/(1 - \cos R)$ ):

$$O(P_\mu) = E - \beta \frac{M^2}{E}.$$

- ▶ This is a global optimization problem!
- ▶ Quantum anneal using modified QUBO objective function:

$$O'(P_\mu) = E^2 - \beta M^2 = \sum_{i,j=1}^N \beta \left( \vec{p}_i \cdot \vec{p}_j + \frac{1-\beta}{\beta} E_i E_j \right) x_i x_j.$$

# Multi-region Optimization

Simultaneously find  $M$  jets.

- ▶  $O(N(M + 1))$  qubits  $x_{ir}$ ,  $i \in \{1, 2, \dots, N\}$ ,  $r \in \{1, 2, \dots, M\}$ .
- ▶ Generalize modified jet function objective function

$$O(\{x_{ij}\}) = \sum_{r=1}^M \sum_{i,j=1}^N \beta \left( \vec{p}_i \cdot \vec{p}_j + \frac{1-\beta}{\beta} E_i E_j \right) x_{ir} x_{jr} \\ + \Lambda \sum_{i=1}^N \left( 1 - \sum_{r=0}^M x_{ir} \right)^2 .$$

- ▶ Last penalty term ensures each particle is only assigned to one region.



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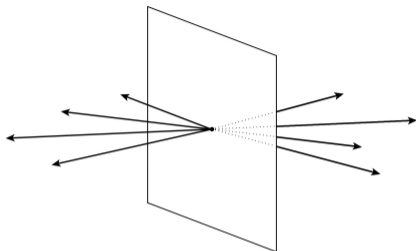
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# Conclusions



- ▶ Explored how to realistically apply quantum computing to collider physics problems.
  - ▶ In particular, how do we handle data loading in the quantum gate model?
  - ▶ Improved classical  $O(N^3)$  thrust algorithm to  $O(N^2 \log N)$  classical algorithm and  $O(N^2)$  quantum algorithm.
- ▶ Considerations are relevant to broader optimization and clustering problems.

## Future Directions

- ▶ Can we identify other interesting problems from collider physics where the cost of data read-in is  $O(N)$ , while the size of the search space goes like  $O(N^\alpha)$  for large  $\alpha$ ?