

# Quantum search of partially ordered sets

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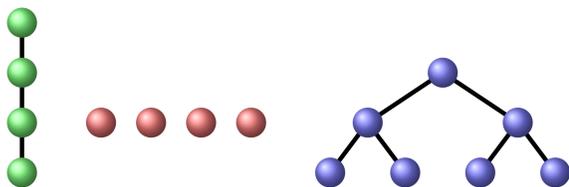
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## Abstract

- We investigate the generalisation of quantum search of unstructured and totally ordered sets to search of partially ordered sets (posets).
- In two models, we show that quantum algorithms can achieve at most a quadratic improvement in query complexity over classical algorithms, up to logarithmic factors; we also give quantum algorithms that almost achieve this bound.
- In one model, we give an almost optimal quantum algorithm for searching forest-like posets.
- In the other, we give an optimal  $O(\sqrt{n})$  quantum algorithm for searching posets derived from  $n \times n$  arrays sorted along rows and columns.
- This leads to an optimal  $O(\sqrt{n})$  quantum algorithm for finding the intersection of two sorted lists of  $n$  integers.

## Partially ordered sets

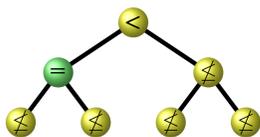
- A **partial order** on a set  $S$  is a relation  $\leq$  such that, for  $a, b, c \in S$ ,  $a \leq a$ ,  $(a \leq b) \wedge (b \leq a) \Rightarrow a = b$ , and  $(a \leq b) \wedge (b \leq c) \Rightarrow a \leq c$ .
- Posets can be expressed by **Hasse diagrams**:



## Two models for search

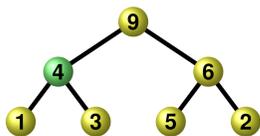
In the **abstract** model [4]:

- We are searching for an unknown “marked” element  $a$ .
- Querying element  $x$  returns  $<$ ,  $=$ ,  $\neq$  depending on whether  $a < x$ ,  $a = x$ , or either  $a > x$  or  $a$  and  $x$  are incomparable.



In the **concrete** model [6]:

- Each element  $s \in S$  stores an unknown integer  $S[s]$ .
- For any given integer  $a$ , we want to find the unique  $s$  such that  $S[s] = a$ .



## General bounds

**Theorem.** Let  $S$  be an  $n$ -element poset, and let  $D(S)$ ,  $Q_E(S)$  and  $Q_2(S)$  be the number of queries required for an exact classical, exact quantum, or bounded-error quantum (respectively) algorithm to find the marked element in  $S$ . Then, in the abstract model,

$$D(S) = O(Q_2(S)^2 \log n)$$

$$Q_2(S) = O(\sqrt{D(S)} \log n \sqrt{\log \log n})$$

and in the concrete model,

$$D(S) = O(Q_2(S)^2 \log n)$$

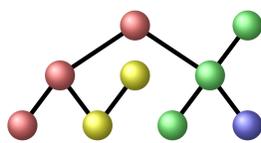
$$Q_E(S) = O(\sqrt{D(S)} \log n)$$

Proof idea (abstract model):

- Reduce poset search to an **oracle identification** problem (finding marked element  $\Leftrightarrow$  identifying an oracle).
- Lower bound: from a result of Servedio and Gortler [7]. Upper bound: from a result of Atici and Servedio [3].

Proof idea (concrete model):

- Lower bound: from the bound of Ambainis on inverting a permutation [2].
- Upper bound: from **Dilworth's Theorem** [5] giving a decomposition of posets into **chains** (sets of comparable elements). Perform binary search on each chain in quantum parallel.



## Recursive quantum search

**Theorem.** Let  $P_n$  be the problem of searching an abstract database, parametrised by an abstract size  $n$ , for a known element which may or may not be in the database. Let  $T(n)$  be the time required for a bounded-error quantum algorithm to solve  $P_n$ , i.e. to find the element, or output “not found”. Let  $P_n$  satisfy the following conditions:

- If  $n \leq n_0$  for some constant  $n_0$ , then there exists an algorithm to find the element, if it is contained in the database, in time  $T(n) \leq t_0$ , for some constant  $t_0$ .
  - If  $n > n_0$ , then the database can be divided into  $k$  sub-databases of size at most  $\lceil n/k \rceil$ , for some constant  $k > 1$ .
  - If the element is contained in the original database, then it is contained in exactly one of these sub-databases.
  - Each division into sub-databases uses time  $f(n)$ , where  $f(n) = O(n^{1/2-\epsilon})$  for some  $\epsilon > 0$ .
- Then  $T(n) = O(\sqrt{n})$ .

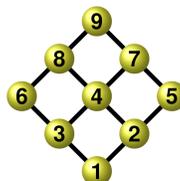
Proof idea:

- Based on a result of Aaronson and Ambainis [1] on quantum search of spatial regions.
- Split the database some number of times and pick a sub-database at random, then recurse.
- Perform some number of iterations of amplitude amplification at each recursive step...
- ...then use amplitude amplification on the whole algorithm.

## 2-dimensional arrays

An interesting poset: an  $n \times n$  array of distinct integers that are increasing along rows and columns.

1	3	6
2	4	8
5	7	9



- We give an optimal quantum algorithm which finds an integer in such an array using  $O(\sqrt{n})$  queries.
- Implies an optimal  $O(n^{(d-1)/2})$  algorithm to find an integer in a  $d$ -dimensional  $n \times n \times \dots \times n$  array.

- Also implies an optimal  $O(\sqrt{n})$  algorithm to find the intersection of two sorted lists of  $n$  integers.

- The quantum algorithm is based on an asymptotically optimal  $O(n)$  recursive classical algorithm.

- We use the **recursive quantum search theorem** above to convert this to an optimal quantum algorithm.

Sketch of the classical algorithm:

- Perform binary search on the central row (or column) of the array.

- Can discard at least half the elements, leaving two rectangular subarrays.

- Call this algorithm recursively on these subarrays.

**Example:** (where green: integer to search for, yellow: not searched yet, blue: currently being searched, red: discarded)

1	3	5	10	13
2	4	7	11	14
6	8	9	15	21
12	16	17	20	24
18	19	22	23	25

1	3	5	10	13
2	4	7	11	14
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12	16	17	20	24
18	19	22	23	25

## Conclusions

- We have given general upper and lower bounds on quantum search of partially ordered sets, in two different models.

- The non-query transformations used by the algorithms given here are efficiently implementable.

- Given a poset  $S$  to be searched, quantum circuits for these algorithms can be produced in time polynomial in the size of  $S$ .

Open questions:

- In the abstract model, is there a general lower bound of  $Q_2(S) = \Omega(\log n)$ ?

- Can the logarithmic factors in the quantum upper bounds in both models be improved, e.g. by being changed into additive terms?

- In the concrete model, could the 2D search algorithm be extended to arrays that may contain duplicate elements?

## References

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