

# Upper bounds for query complexity inspired by the Elitzur-Vaidman bomb tester

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# Overview

## 1 Bomb Query Complexity

- Elitzur-Vaidman bomb tester
- Bomb query complexity  $B(f)$
- Main result:  $B(f) = \Theta(Q(f)^2)$

## 2 Algorithms

- Introduction:  $O(N)$  bomb query algorithm for  $OR$
- Main theorem 2: constructing q. algorithms from c. ones
- Applications: graph problems

## 3 Summary and open problems

# Section 1

## Bomb Query Complexity

# Elitzur-Vaidman Bomb Tester [EV93]

A collection of bombs, some of which are duds

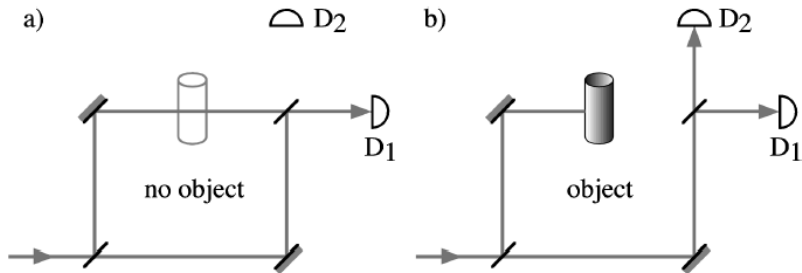
Live: Explodes on contact with photon

Dud: No interaction with photon

Can we tell them apart without blowing ourselves up?

# Elitzur-Vaidman Bomb Tester [EV93]

We can put a bomb in an Mach-Zehnder interferometer:

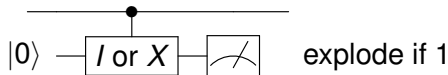


If  $D_2$  detects a photon, then we know the bomb is live, even though it has not exploded.

Image source: A. G. White et al., PRA 58, 605 (1998).

# EV bomb in circuit model

We can rewrite the Elitzur-Vaidman bomb in the circuit model:

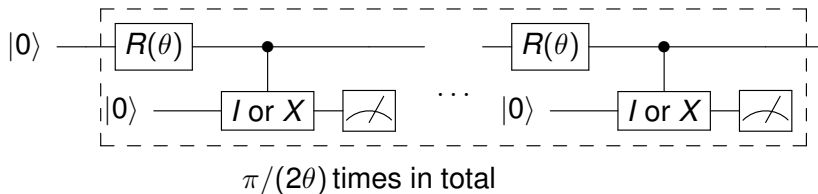


Live bomb:  $X$  in the above diagram

Dud:  $I$  in the above diagram

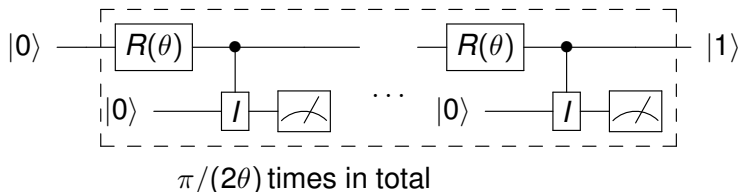
# Quantum Zeno Effect [KWH+95]

Let  $R(\theta) = \exp(i\theta X) = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$ .



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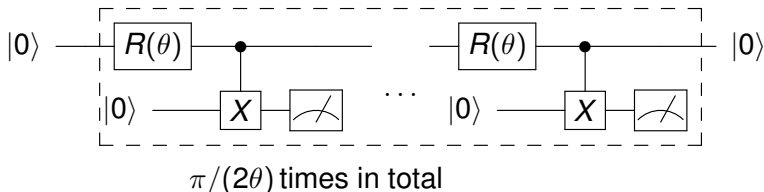


If dud: Ctrl- $I$  does nothing, so  $|0\rangle$  gets rotated to  $|1\rangle$ .



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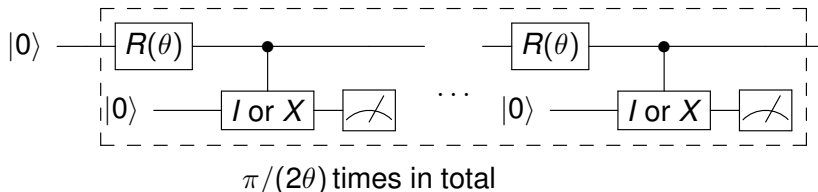


If live: First register is projected back to  $|0\rangle$  on each measurement.

Probability of explosion:  $\Theta(\theta^2) \times \Theta(1/\theta) = \Theta(\theta)$ .

# Quantum Zeno Effect [KWH+95]

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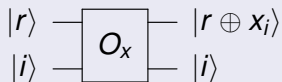


Probability of explosion:  $\Theta(\theta)$

Number of queries:  $\Theta(1/\theta)$

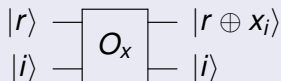
# Quantum Query

## Quantum query

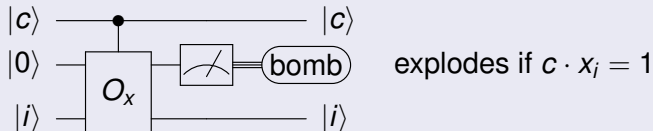


# Quantum Query vs Bomb Query

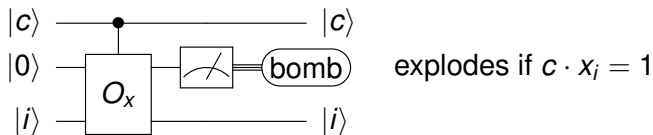
## Quantum query



## Bomb query



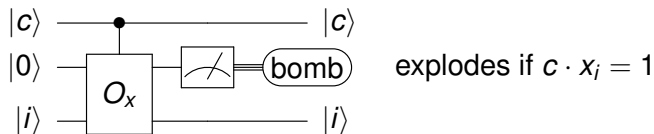
# Bomb Query



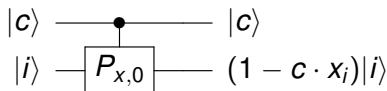
Differences from quantum query:

- Extra control register  $c$ .
- The record register, where we store the query result, *must* contain 0 as input.
- We *must* measure the query result after each query; if the result is 1, the bomb explodes and the algorithm fails.

# Bomb Query



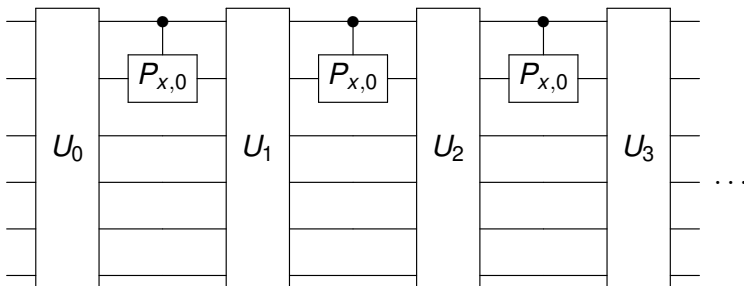
equivalent to



where

$$P_{x,0} = \sum_{x_i=0} |i\rangle\langle i|, \quad \text{Ctrl} - P_{x,0} = I - \sum_{x_i=1} |1, i\rangle\langle 1, i|$$

# Bomb Query Complexity



Call the minimum number of bomb queries needed to determine  $f$  with bounded error, with probability of explosion  $\leq \epsilon$ , the bomb query complexity  $B_\epsilon(f)$ .

# Main Theorem

## Theorem

$$B_{\epsilon}(f) = \Theta(Q(f)^2/\epsilon).$$

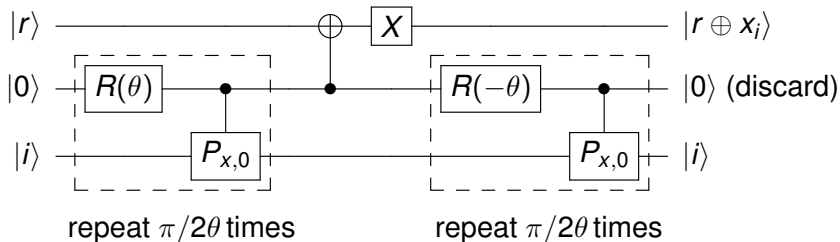
Upper bound: Quantum Zeno effect.

Lower bound: Adversary method.



# $B_\epsilon(f) = O(Q(f)^2/\epsilon)$ : Proof

We can simulate each quantum query using  $\Theta(1/\theta)$  bomb queries:



Total probability of explosion:  $\Theta(\theta) \cdot Q(f) = \Theta(\epsilon)$ , if  $\theta = \Theta(\epsilon/Q(f))$ .

Total number of bomb queries:  $\Theta(1/\theta) \cdot Q(f) = O(Q(f)^2/\epsilon)$ .

# $B_\epsilon(f) = \Omega(Q(f)^2/\epsilon)$ : Proof

The proof uses the general-weight adversary method [HLS07]. We know [Rei09, Rei11, LMR+11] that the general-weight adversary bound tightly characterizes quantum query complexity:  $\text{Adv}^\pm(f) = \Theta(Q(f))$ .

By modifying the proof of the general-weight adversary bound, we can show that  $B_\epsilon(f) = \Omega(\text{Adv}^\pm(f)^2/\epsilon)$ .

This implies that  $B_\epsilon(f) = \Omega(Q(f)^2/\epsilon)$ .

## Section 2

# Algorithms

# $O(N)$ Bomb Query Algorithm for OR

There are  $N$  bombs, want to check if any are live.

Check each bomb using  $\Theta(\epsilon^{-1})$  queries, or  $O(N/\epsilon)$  queries in total.

Each live bomb has  $\Theta(\epsilon)$  chance of exploding.

Each dud has no chance of exploding.

Since we can stop at the first live bomb, the total chance of failure is only  $\Theta(\epsilon)$ . Therefore  $B_\epsilon(OR) = O(N/\epsilon)$ .

Since  $B(OR) = O(N)$ ,  $Q(OR) = O(\sqrt{N})$ .

This is a nonconstructive proof of the existence of Grover's algorithm!

Can we generalize this further?

# Main Theorem 2

## Theorem

*Suppose there is a classical randomized algorithm  $\mathcal{A}$  that computes  $f(x)$  using at most  $T$  queries. Moreover, suppose there is an algorithm  $\mathcal{G}$  that predicts the results of each query  $\mathcal{A}$  makes (0 or 1), making at most an expected  $G$  mistakes.*

*Then  $B(f) = O(TG)$ , and  $Q(f) = O(\sqrt{TG})$ .*

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*Then  $B(f) = O(TG)$ , and  $Q(f) = O(\sqrt{TG})$ .*

For example, for OR we have  $T = N$  and  $G = 1$ , so  $Q(f) = O(\sqrt{N})$ .

# Bomb algorithm with $B(f) = O(TG)$

For each classical query, check whether  $\mathcal{G}$  correctly predicts the query result of  $\mathcal{A}$  using  $\Theta(G/\epsilon)$  bomb queries.

If  $\mathcal{G}$  guesses incorrectly then the probability of explosion is  $O(\epsilon/G)$ ; otherwise it is zero. (This actually requires defining an equivalent symmetric variant of the bomb query complexity.)

The total probability of explosion is  $O(\epsilon/G) \cdot G = O(\epsilon)$ , and the number of bomb queries used is  $O(G/\epsilon) \cdot T = O(TG/\epsilon)$ .



# Explicit q. algorithm with $Q(f) = O(\sqrt{TG})$

- Repeat until all queries of  $\mathcal{A}$  are determined:
  - 1 Use  $\mathcal{G}$  to predict all remaining queries of  $\mathcal{A}$ , under assumption it makes no mistakes.
  - 2 Search for the location  $d_j$  of first mistake, using  $O(\sqrt{d_j - d_{j-1}})$  quantum queries.
  - 3 This determines the actual query results up to the  $d_j$ -th query that  $\mathcal{A}$  would have made.

Kothari's algorithm for oracle identification [Kot14] actually already uses these steps above.

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$i$	2	5	4	3	12	7	6	9	10
$x_i$	0	1	1	0	1	1	0	0	1

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Query complexity:  $O(G) \cdot O(\sqrt{T/G}) = O(\sqrt{TG})$ .

It looks like error reduction may give extra log factors, but [Kot14] showed that the log factors can be removed using span programs.



# Applications: Breadth First Search

## Problem: Unweighted Single-Source Shortest Paths

Given the adjacency matrix of an unweighted graph as a black box, find the distances from a vertex  $s$  to all other vertices.

Classical algorithm: *Breadth First Search*.

## Breadth First Search

- 1 Initialize an array  $dist$  that will hold the distances of the vertices from  $s$ . Set  $dist[s] := 0$ , and  $dist[v] := \infty$  for  $v \neq s$ .
- 2 For  $d = 1, \dots, n - 1$ :
  - 1 For all vertices  $v$  with  $dist[v] = d - 1$ , query its outgoing edges  $(v, w)$  to all vertices  $w$  whose distance we don't know ( $dist[w] = \infty$ ). If  $(v, w)$  is an edge, set  $dist[w] := d$ .

# BFS: Quantum Query Complexity

## Breadth First Search

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Worst case query complexity is  $T = O(n^2)$ , where  $n$  is no. of vertices. If we guess that each queried pair  $(v, w)$  is not an edge, then we make at most  $G = n - 1$  mistakes, since each vertex is only discovered once.

$Q(uSSSP) = O(\sqrt{TG}) = O(n^{3/2})$ , matches lower bound of [DHH+04].

# Applications: $k$ -Source Shortest Paths

What if we instead want the distances from  $k$  different sources?

## Problem: Unweighted $k$ -Source Shortest Paths

Given the adjacency matrix of an unweighted graph as a black box, find the distances from vertices  $s_1, \dots, s_k$  to all other vertices.

Classical: Run BFS  $k$  times.

Quantum:  $G = k(n - 1)$ , but  $T = O(n^2)$  instead of  $O(kn^2)$ .  
Therefore  $Q(kSSP) = O(k^{1/2}n^{3/2})$ .

Dhariwal and Mayar showed tight lower bound;  
available on S. Aaronson's blog, Dec. 26, 2014:

<http://www.scottaaronson.com/blog/?p=2109>

# Applications: Maximum Bipartite Matching

## Problem: Maximum Bipartite Matching

A *matching* in an undirected graph is a set of edges that do not share vertices. Given a bipartite graph, find a matching with the maximum possible number of edges.

Classical algorithm: Hopcroft-Karp algorithm.

Essentially proceeds by using  $O(\sqrt{n})$  rounds of BFS and modified DFS (depth-first search).

Quantum:  $G = O(\sqrt{n} \times n) = O(n^{3/2})$ , and  $T = O(n^2)$  (not  $O(n^{2.5})$ ).  
Therefore  $Q(MBM) = O(n^{7/4})$ . First nontrivial upper bound!

# Summary

- Inspired by the EV bomb tester, we defined the notion of *bomb query complexity*, and showed the relation  $B(f) = \Theta(Q(f)^2)$ .
- Bomb query complexity further lead us to a general construction of quantum query algorithms from classical algorithms, giving us an  $O(n^{1.75})$  quantum query algorithm for maximum bipartite matching.

# Open Questions

- Can we relate  $G$ , the number of wrong guesses, to classical measures of query complexity (e.g. certificate, sensitivity...)?
- Time complexity of algorithms?
- Algorithms for adjacency list model?
- Other problems e.g. matching for general graphs?
- Relationship between  $R(f)$  and  $B(f)$ ?

# Relationship between $R(f)$ and $B(f)$ ?

For total functions the largest known separation between  $R(f)$  and  $Q(f)$  is quadratic (for the OR function).

It is conjectured this is the extreme case,  $R(f) = O(Q(f)^2)$ .

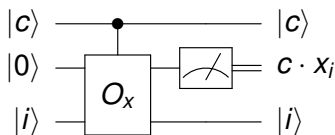
We know that  $B(f) = \Theta(Q(f)^2)$ .

Therefore the conjecture is equivalent to  $R(f) = O(B(f))$ .

We give some motivation for why this conjecture might be true...

# Projective Query Complexity, $P(f)$

Aaronson (unpublished, 2002) considered allowing access to the black box only with the following:



We call the number of queries required the *projective query complexity*,  $P(f)$ . Note the algorithm does *not* end on measuring a 1.

Straightforwardly  $Q(f) \leq P(f) \leq R(f)$  and  $P(f) \leq B(f)$ .

Regev and Schiff [RS08]:  $P(OR) = \Omega(N)$ .

Open question: Does  $P(f) = \Theta(R(f))$  for all total functions?

If this is true, implies  $R(f) = O(B(f)) = O(Q(f)^2)$ .



# Thank You!