

# Optimal quantum adversary lower bounds for ordered search

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**A mathematical question**

## A mathematical question

Given a matrix

$$A_n = \begin{pmatrix} a_0 & a_1 & a_2 & a_3 & \dots & a_{n-1} \\ a_1 & a_2 & a_3 & \dots & a_{n-1} & 0 \\ a_2 & a_3 & \dots & a_{n-1} & 0 & 0 \\ \vdots & \dots & & & \vdots & \vdots \\ a_{n-2} & a_{n-1} & 0 & 0 & 0 & 0 \\ a_{n-1} & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

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how large can  $\sum_i a_i$  be while  $\|A_n\| \leq 1$ ? Let  $\alpha(n)$  denote this optimal value.

## A good guess

The “half” Hilbert matrix

$$Z_n = \begin{pmatrix} 1 & 1/2 & 1/3 & 1/4 & \dots & 1/n \\ 1/2 & 1/3 & 1/4 & \dots & 1/n & 0 \\ 1/3 & 1/4 & \dots & 1/n & 0 & 0 \\ \vdots & \dots & & & \vdots & \vdots \\ 1/(n-1) & 1/n & 0 & 0 & 0 & 0 \\ 1/n & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

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Then  $\sum_i a_i \approx \ln(n)$ . How to upper bound  $\|Z_n\|$ ?

## Hilbert's Inequality

Consider the Hilbert matrix

$$H = \begin{pmatrix} 1 & 1/2 & 1/3 & 1/4 & \dots \\ 1/2 & 1/3 & 1/4 & \dots & \dots \\ 1/3 & 1/4 & \dots & & \dots \\ 1/4 & \dots & & & \vdots \\ \vdots & & & \vdots & \ddots \end{pmatrix}$$

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Hilbert showed (with improvement by Schur) that  $\|H\| \leq \pi$ . Thus the (normalized) half Hilbert matrix demonstrates  $\alpha(n) \geq \frac{\ln(n)}{\pi}$ .

## Our main theorem

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$$\alpha^+(n) = \sum_{i=0}^{n-1} \left( \frac{\binom{2i}{i}}{4^i} \right)^2 = \frac{1}{\pi} (\ln n + \gamma + \ln 8) + O(1/n)$$

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Note that

$$\frac{\binom{2i}{i}}{4^i} \approx \frac{4^i / \sqrt{\pi i}}{4^i} = \frac{1}{\sqrt{\pi i}}.$$

## Motivation: quantum query complexity

- In classical query complexity, want to compute some function  $f(x)$  and have access to the input  $x$  by queries of the form  $x_i = ?$ . Complexity is number of queries needed on worst case input.
- Model of quantum query complexity is attractive as captures many quantum algorithms
  - Grover's search algorithm,
  - Period finding of Shor's algorithm,
  - Quantum walks: element distinctness, triangle finding, matrix multiplication
- And we can also prove lower bounds!

## Ordered search problem

- Complexity of finding a given item in an ordered list.
- Given an ordered list  $x_1 \leq x_2 \leq \dots \leq x_n$  want to find position of given item  $z$ .
- Ask queries of the form  $x_i \geq z$ ?
- Equivalently can represent problem as querying bits of input and identifying first occurrence of a '1'. For  $n = 4$ , for example  $S = \{1111, 0111, 0011, 0001\}$ .

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- What is this fundamental constant of quantum information?

## Lower bounds: adversary method

- Main lower bound techniques: polynomial method and adversary method.
- Adversary method developed and improved in long series of works [BBBV94, Amb00, HNS01, BSS03, Amb03, LM04, Zha04, SŠ06, HLŠ07]
- Adversary bound is an optimization problem which can be written as a semidefinite program.

$$\text{ADV}(f) := \max_{\Gamma} \frac{\|\Gamma\|}{\max_i \|\Gamma \circ D_i\|}$$

where  $\Gamma[x, y] = 0$  if  $f(x) = f(y)$  and  $D_i[x, y] = 1$  if  $x_i \neq y_i$  and 0 otherwise.

## The $\Gamma$ matrix

	$f^{-1}(0)$	$f^{-1}(1)$
$f^{-1}(0)$	0	$A$
$f^{-1}(1)$	$A^*$	0

Notice that the spectral norm of  $\Gamma$  equals that of  $A$ .

## The $\Gamma \circ D_1$ matrix

		$f^{-1}(0)$		$f^{-1}(1)$	
		$0x$	$1x$	$0y$	$1y$
$f^{-1}(0)$	$0x$	0		0	<b>B</b>
	$1x$			<b>C</b>	0
$f^{-1}(1)$	$0y$	0	<b>C*</b>	0	
	$1y$	<b>B*</b>	0		

The spectral norm of  $\Gamma \circ D_1$  equals  $\max\{\|B\|, \|C\|\}$ .

## Automorphism principle

- “Whenever you have to deal with a structure endowed entity  $\Sigma$  try to determine its group of automorphisms . . . you can expect to gain a deep insight into the constitution of  $\Sigma$  in this way.”

—Hermann Weyl, *Symmetry*

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- Using symmetry of problem can greatly simplify search for optimal adversary matrices [HLŠ07].
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- Input to ordered search (for  $n = 4$ )  $S = \{1111, 0111, 0011, 0001\}$  Trivial automorphism group!

## Automorphism principle

- [FGGS99] extend inputs “to a circle”:  $S' = \{11110000, 01111000, 00111100, 00011110, 00001111, 10000111, 11000011, 11100001\}$
- Now have cyclic structure, and query complexity changes by at most 1.
- Using automorphism principle, can wlog reduce computation of adversary bound to the matrix problem given at beginning of talk.

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- Now have cyclic structure, and query complexity changes by at most 1.
- Using automorphism principle, can wlog reduce computation of adversary bound to the matrix problem given at beginning of talk.
- We show that the adversary method (even with negative weights) cannot show lower bounds larger than  $\frac{1}{\pi} \ln n + O(1)$ .

## A word about the proof (non-negative case)

- We exhibit solutions to both the primal and dual formulation of adversary bound, and show that they match.
- A key role in both directions is played by the lovely sequence

$$\beta_i = \frac{\binom{2i}{i}}{4^i}.$$

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- Key property:  $\sum_{i=0}^j \beta_i \beta_{j-i} = 1$

- Proof:

$$\frac{1}{\sqrt{1-z}} = \beta_0 + \beta_1 z + \beta_2 z^2 + \beta_3 z^3 + \dots$$

## Optimal matrix

Recall we wish to show that  $\alpha^+(n) = \sum_{i=0}^{n-1} \left( \frac{\binom{2i}{i}}{4^i} \right)^2$ .

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Define  $A_n(j) = \sum_{i=0}^{n-j-1} \beta_i \beta_{i+j}$ .

$$\begin{pmatrix} A_4(0) - A_4(1) & A_4(1) - A_4(2) & A_4(2) - A_4(3) & A_4(3) \\ A_4(1) - A_4(2) & A_4(2) - A_4(3) & A_4(3) & 0 \\ A_4(2) - A_4(3) & A_4(3) & 0 & 0 \\ A_4(3) & 0 & 0 & 0 \end{pmatrix}$$

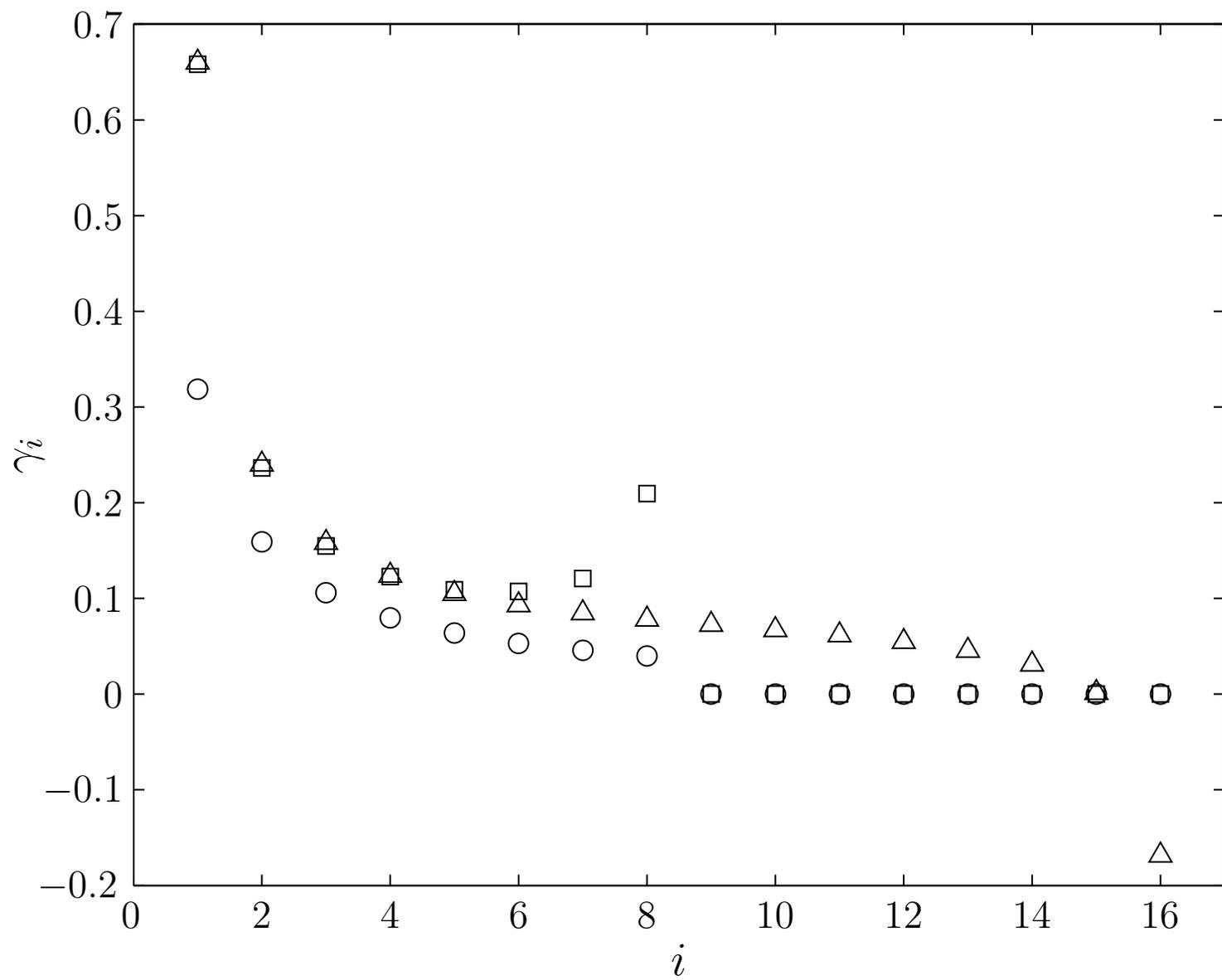
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To bound spectral norm, show that  $x = [\beta_3, \beta_2, \beta_1, \beta_0]$  is eigenvector with eigenvalue 1.



## Conclusion

- What is the quantum query complexity of ordered search?
- Progress will require new algorithms or new lower bound techniques.
- [BSS03] show quantum query complexity can be written as a semidefinite program. Adversary bound can be viewed as a relaxation of this program.
- Our optimal matrix can be used to give nearly elementary proof of Hilbert's Inequality (need  $\Gamma(1/2) = \sqrt{\pi}$ ).