Quantum query complexity

Lecture 4 The adversary method

Materials: https://yassine-hamoudi.github.io/pcmi2023/

Focus of this lecture

- A lower bound method that is always optimal > We'll show in lecture 5 how to turn it into an algorithm > Counterpart: often harder to use
- It shares some ideas with the hybrid method (lecture 1) and the recording method (lecture 3) (in fact: these can be seen as particular cases of it)

The (generalized) adversary method

The distinguishing lemma (lecture 1)

The purification viewpoint (lecture 3)

$$\sum_{x} \sqrt{p_{x}} |x\rangle$$

$$\begin{vmatrix} 0 \\ |0 \rangle \end{vmatrix} U_{0}$$

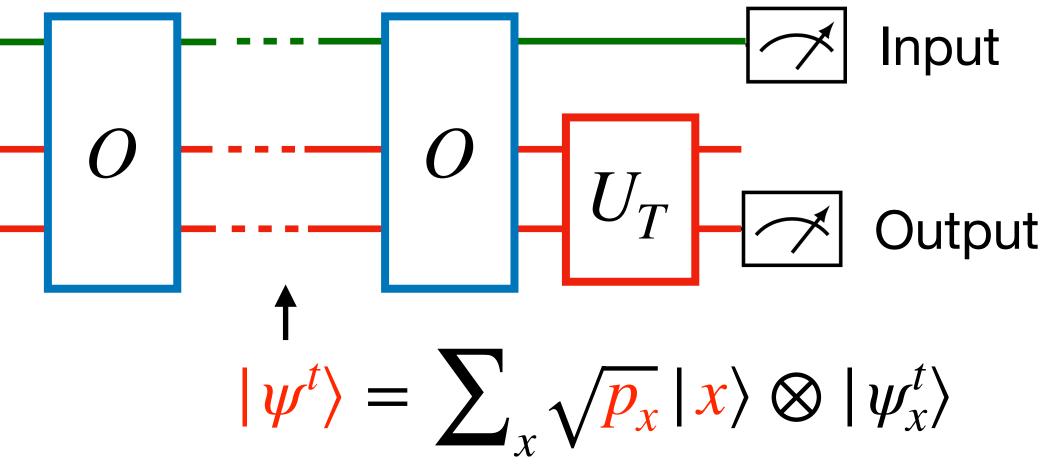
$$O$$

$$U_{1}$$

Reminders

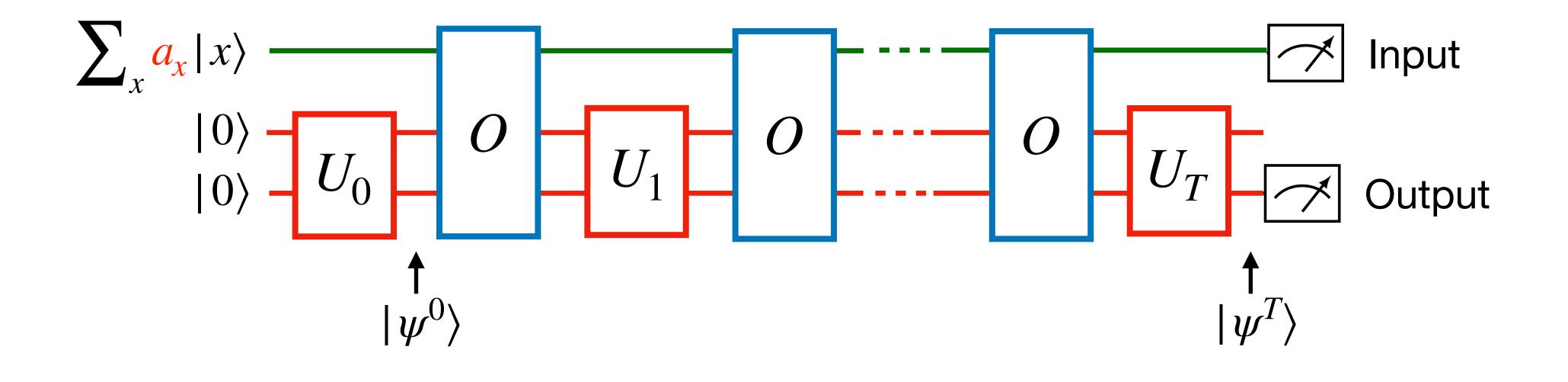
The states $|\psi_x^T\rangle$ and $|\psi_y^T\rangle$ can be distinguished with probability $\geq 2/3$ if an only if there are "sufficiently orthogonal" $|\langle \psi_x^T | \psi_v^T \rangle| \le 2\sqrt{2/3}$

We can set a distribution $(p_x)_x$ on the input by adding a purification register





Quantum adversary



<u>First step:</u> replace $(p_x)_x$ with complex numbers $(a_x)_x$ s.t. $\sum_x |a_x|^2 = 1$

$|\psi^t\rangle = \sum_x a_x |x\rangle \otimes |\psi^t_x\rangle$

<u>Second step:</u> consider the Gram matrix: $\sum_{x,y} a_x^* a_y$

<u>Third step</u>: place some weights $\Gamma_{x,y}$ on the "hard" pairs of inputs (symmetric) $\Gamma_{x,v} = \Gamma_{v,x}$ $x \left| \cdots \Gamma_{x,y} a_x^* a_y \langle \psi_x^t | \psi_y^t \rangle \right|$

if f(x) = f(y) then $\Gamma_{x,y} = 0$ (consistent)

<u>First step:</u> replace $(p_x)_x$ with complex numbers $(a_x)_x$ s.t. $\sum_x |a_x|^2 = 1$

 $|\psi^t\rangle = \sum_{x} a_x |x\rangle \otimes |\psi_x^t\rangle$

$$\langle \psi_x^t | \psi_y^t \rangle | x \rangle \langle y |$$

 $x \left(- - - - a_x^* a_y \langle \psi_x^t | \psi_y^t \rangle \right)$



Adversary matrix: $\Gamma \in \mathbb{R}^{2^n \times 2^n}$ symmetric and $f(x) = f(y) \Rightarrow \Gamma_{x,v} = 0$

Adversary distribution: $a \in \mathbb{C}^{2^n}$ principal (unit) eigenvector of Γ

" **"Punctured" matrices:** $\Gamma_i \in \mathbb{R}^{2^n}$

Progress measure: $\Delta_t = |\langle \psi^t | (\mathbf{I}) \rangle$

<u>Lemma 1:</u> $\Delta_0 = |\Gamma|$

Lemma 2: $\Delta_T < 0.95 \|\Gamma\|$ if the algorithm succeeds wp $\geq 2/3$

 $\underline{\text{Lemma 3: }} \Delta_{t+1} \ge \Delta_t - 2 \max_{1 \le i \le n} \|\Gamma_i\|$

^{×2ⁿ} such that
$$(\Gamma_i)_{x,y} = \Gamma_{x,y} \cdot \mathbf{1}_{x_i \neq y_i}$$

 $\Gamma \otimes \mathrm{Id} |\psi^t\rangle| = \Big|\sum_{x,y} \Gamma_{x,y} a_x^* a_y \langle \psi_x^t |\psi_y^t\rangle$
(initial condi

(final condition)

(evolution)



Theorem: $Q(f) \ge$

- Positive-weight adversary: $\forall x$,
 - Has a nice combinatorial interpretation (see problem session)
 - Sub-optimal (the "certificate" and "property testing" barriers)

- Negative-weight adversary: $\forall x$,
 - Optimal! (see next lecture) Q(f)

$$\geq \max_{\Gamma} \frac{\|\Gamma\|}{40 \cdot \max_{1 \leq i \leq n} \|\Gamma_i\|}$$

$$y, \Gamma_{x,y} \ge 0$$

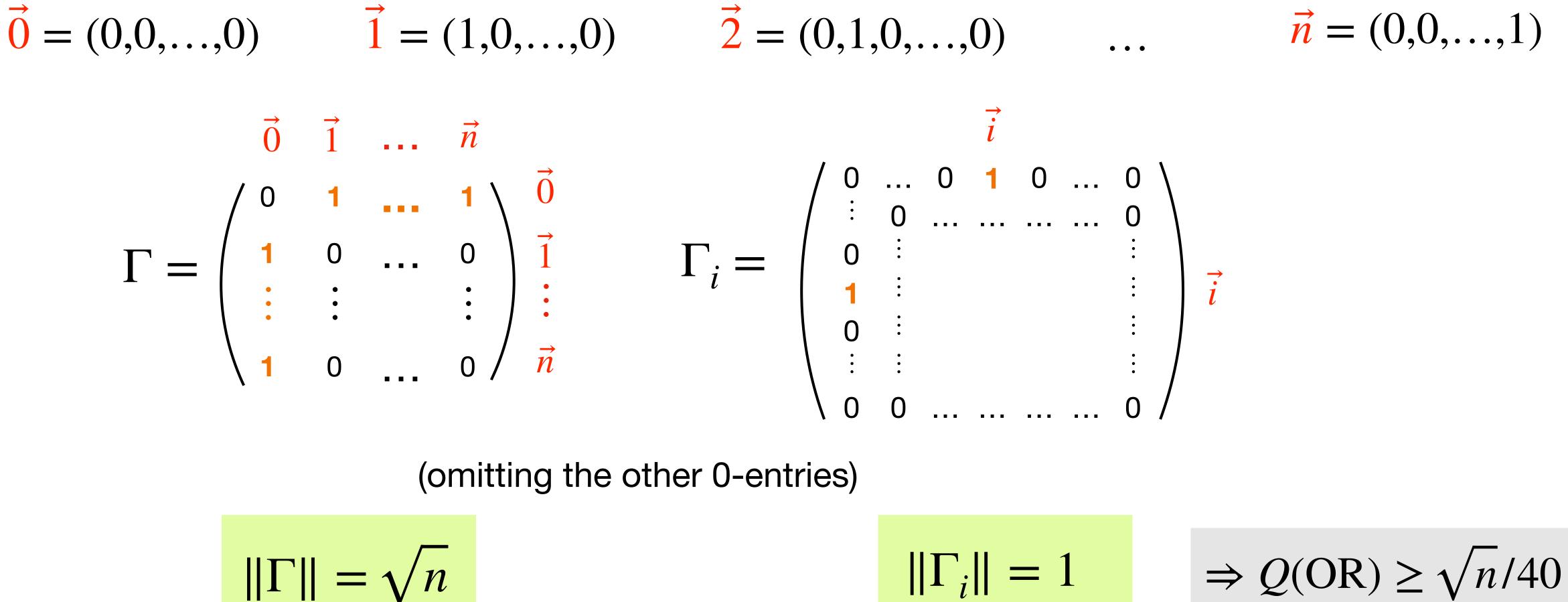
etation (see problem session) d "property testing" barriers)

$$= \Theta\left(\max_{\Gamma} \frac{\|\Gamma\|}{\max_{1 \le i \le n} \|\Gamma_i\|}\right)$$

Applications

<u>OR function:</u> f(x) = 0 if and only if $x = (0,0,\ldots,0)$

We (again) only focus on the n + 1 "hardest" inputs denoted by:



$$\|\Gamma\| = \sqrt{n}$$