

# Quantum PRP/PRF Switching Lemma via Adversary Method



# Towards Tight Bounds for 3-Collision

### Ansis Rosmanis

Graduate School of Mathematics, Nagoya University

# k-Collision problem

Given an oracle access to  $f: \{1, \ldots, n\} \rightarrow \{1, \ldots, n\}$ ,

$$x \longrightarrow O_f \longrightarrow f(x)$$

decide if there are k distinct  $x_1, x_2, \ldots, x_k$  such that  $f(x_1) = f(x_2) = \ldots = f(x_k)$ .

#### Query complexity

$$k = 2 : \Theta(n^{2/3}),$$
  
 $k = 3 : \Omega(n^{2/3}), O(n^{5/7}),$   
 $k = 4 : \Omega(n^{11/16}), O(n^{11/15}).$ 

## Importance

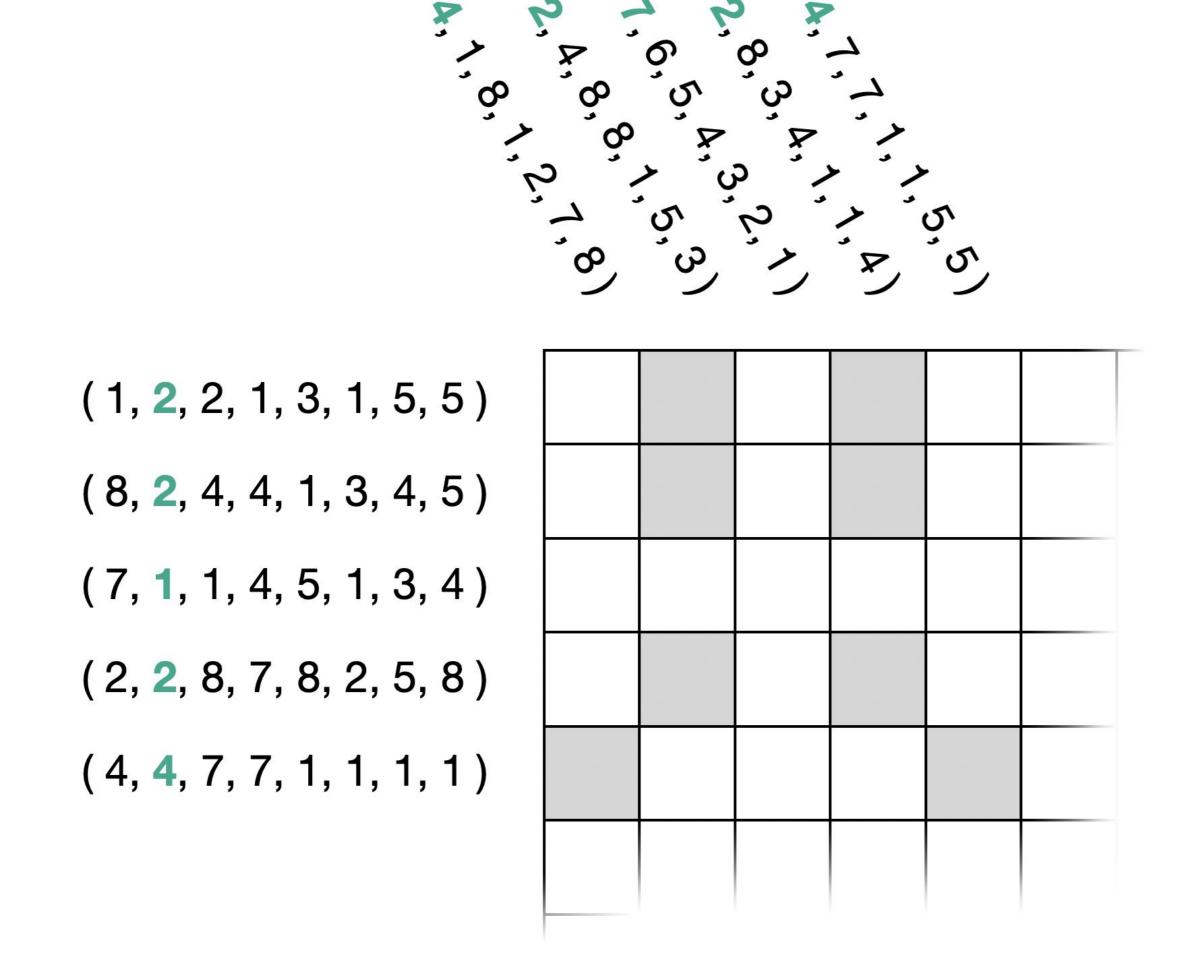
- New algorithmic techniques,
- New lower bound techniques,
- Cryptographic applications.

#### Quantum adversary method

For a decision problem P, one has to decide if  $f \in Yes_P$  or  $f \in No_P$ .

$$f = (f(1), f(2), \dots, f(n))$$

 $\Gamma$  – any non-zero  $Yes_P \times No_P$  matrix.



For every input  $x \in \{1, \dots, n\}$ ,

$$\Gamma_{f,g}^{(x)} := \begin{cases} \Gamma_{f,g} & \text{if } f(x) \neq g(x), \\ 0 & \text{if } f(x) = g(x). \end{cases}$$

# Adversary bound

Bounded-error quantum query complexity

$$Q_{\epsilon}(\mathsf{P}) = \Omega\left(\|\Gamma\|/\max_{x}\|\Gamma^{(x)}\|\right).$$

#### Challenges

- 1. Choosing a good  $\Gamma$ ,
- 2. Evaluating the norms.

#### Hardest instances for 3-Collision

 $No_{3C}$ :  $im_{mult} f$  contains n/2 pairs;

 $Yes_{3C}$ : im<sub>mult.</sub> f contains n/2-2 pairs and a quadruple.

The structure makes constructing lower bounds for the k=3 case more difficult than for the k=2 case.

#### Assignment vectors

Assignment: partial function  $\alpha \colon \{1, \dots, n\} \to \{1, \dots, n\}$ .

Compatibility: f agrees with  $\alpha$  if  $\alpha(x) = f(x)$  for all  $x \in \text{dom } \alpha$ .

For every Yes-certificate  $\alpha$ :

$$|w_{\alpha}\rangle \stackrel{unit}{:=} \sum_{\substack{f \in Yes \\ \alpha \subset f}} |f\rangle \quad \text{and} \quad |v_{\alpha}\rangle \stackrel{unit}{:=} \sum_{\substack{f \in No \\ \alpha \subset f}} |f\rangle.$$

We interpret the state of the adversary being in  $|w_{\alpha}\rangle$  or  $|v_{\alpha}\rangle$  as the algorithm having learnt that f agrees with  $\alpha$ .

This is reminiscent of dual learning graphs.

# Candidate adversary matrix

Inspired by lower bounds for dual adaptive learning graphs, I propose constructing the adversary matrix  $\Gamma$  as a linear combination of

$$L_{i,o} := \left( \sum_{\substack{\alpha \\ |\text{dom } \alpha| = i \\ |\text{im } \alpha| = o}} |w_{\alpha}\rangle\langle v_{\alpha}| \right) \Pi_{i,o}$$

where  $\Pi_{i,o}$  is a certain projector related to the symmetries of the problem:

$$\Gamma := \sum_{i,o} (n^{5/7} - i - n^{1/7}(i - o)) L_{i,o}.$$