A structural theorem for local algorithms with applications to coding, testing and privacy

Marcel Dall'Agnol

University of Warwick

Tom Gur University of Warwick

Oded Lachish Birkbeck, University of London

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Techniques

Main result

**Applications** 

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Techniques

**Applications** 

An efficient transformation from local

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But why?

Sample-based access buys

Privacy

### Sample-based access buys

O(qt log t) queries in general

•  $O(q \log t)$  queries by **reusing samples** 

- Privacy

  - Efficient repetition: running q-query  $L_1, \ldots, L_t$  on the same
- input takes

## $\mathsf{Local} + \mathsf{robust} \ (\mathcal{O}(1) \ \mathsf{queries})$

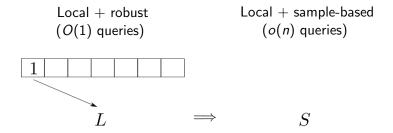
Local + sample-based (o(n) queries)

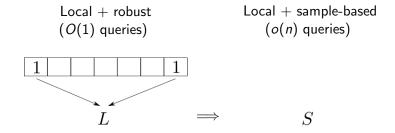
 $L \implies S$ 

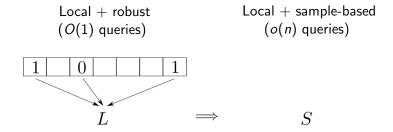
|  |  | eries |  |
|--|--|-------|--|
|  |  |       |  |

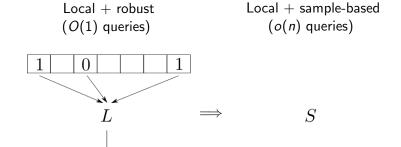
Local | robust

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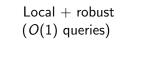


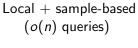


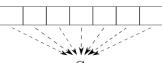
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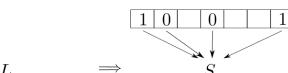




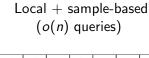


L

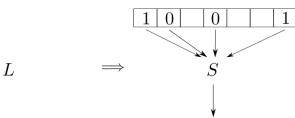
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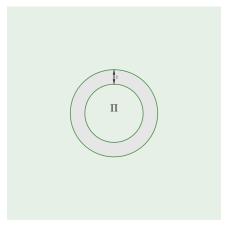
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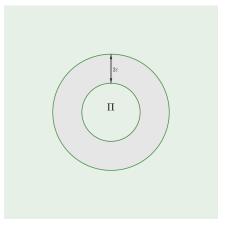
Property  $\Pi \subseteq \{0,1\}^n$ , proximity parameter  $\varepsilon > 0$ 



Tester T computes

$$f(x) = \begin{cases} 1, & \text{if } x \in \Pi \\ 0, & \text{if } x \text{ is } \varepsilon\text{-far from } \Pi. \end{cases}$$

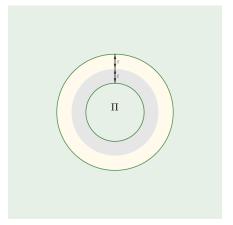
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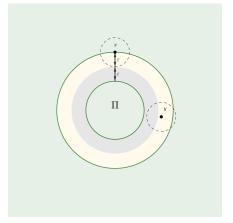
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Tester *T* **robustly** computes

$$g(x) = \begin{cases} 1, & \text{if } x \in \Pi \\ 0, & \text{if } x \text{ is } 2\varepsilon\text{-far from } \Pi. \end{cases}$$

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### Theorem

Any function computed by a q-query robust local algorithm admits a sample-based algorithm with sample complexity

$$n^{1-\tilde{\Omega}(1/q^2)}$$
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$$(q = \Omega(\sqrt{\log n}) \implies \text{sample complexity } \Omega(n))$$

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### **Theorem**

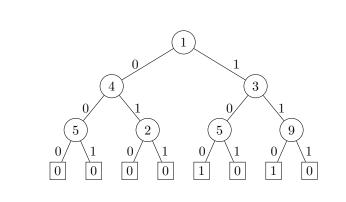
This transformation cannot achieve sample complexity

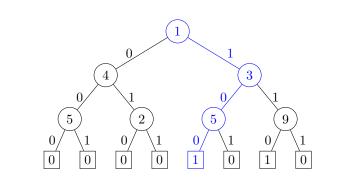
$$n^{1-\omega(1/q)}$$
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### Main result

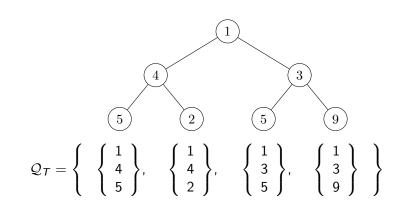
### Techniques

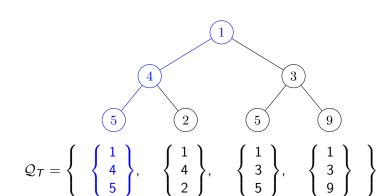
**Applications** 

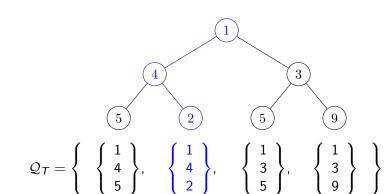


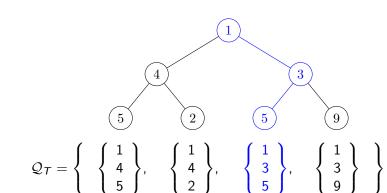


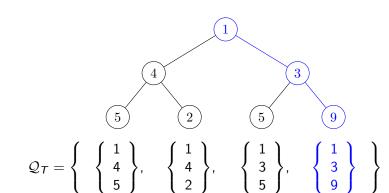
$$x_1 = 1$$
  
 $x_3 = 0$   
 $x_5 = 0$ 











$$\mathcal{Q}_{\mathcal{T}} = \left\{ \begin{array}{c} 1\\4\\5\\5 \end{array}, \quad \left\{ \begin{array}{c}1\\4\\5\\5 \end{array} \right\}, \quad \left\{ \begin{array}{c}1\\4\\2\\5 \end{array} \right\}, \quad \left\{ \begin{array}{c}1\\3\\5\\5 \end{array} \right\}, \quad \left\{ \begin{array}{c}1\\3\\9\\9 \end{array} \right\} \right\}$$

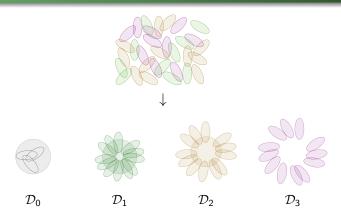
$$Q = \bigcup_T Q_T$$



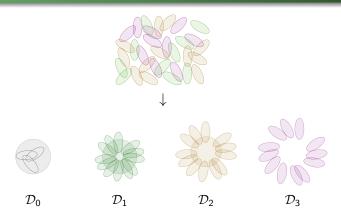
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- small kernel  $|K_i|$
- bounded petal intersection



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Near-optimality of "P  $\neq$  NP for testers"

q-query tester with short proof  $\implies n^{1-\tilde{\Omega}(1/q^2)}$ -query tester.

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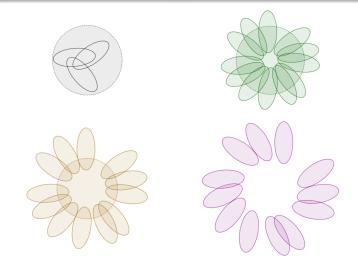
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### Adaptive-to-sample-based transformation for testers

q-query tester  $\implies n^{1-\tilde{\Omega}(1/q^2)}$ -query sample-based tester.



Thank you!