

Property testing against adversarial erasures

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Joint work with Kashyap Dixit, Sofya Raskhodnikova and Abhradeep Thakurta.
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Talk outline

- Property testing

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- Property testing
- Erasure-resilient property testing

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- Erasure-resilient property testing
- Overview of results

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- Overview of results
- Erasure-resilient monotonicity tester
- Current and future directions

Property testing

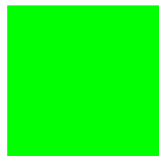
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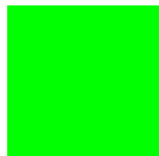


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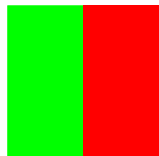
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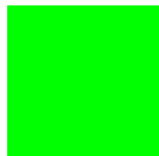
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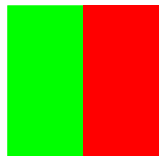
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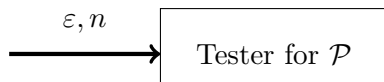
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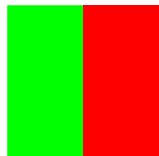
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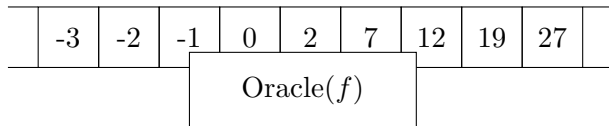
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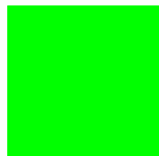
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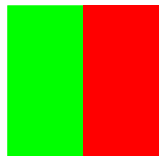
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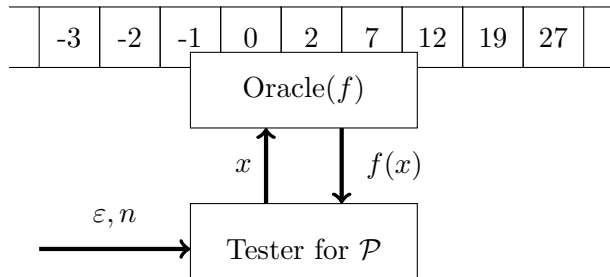


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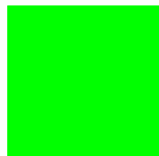
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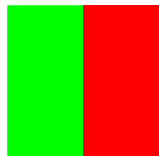
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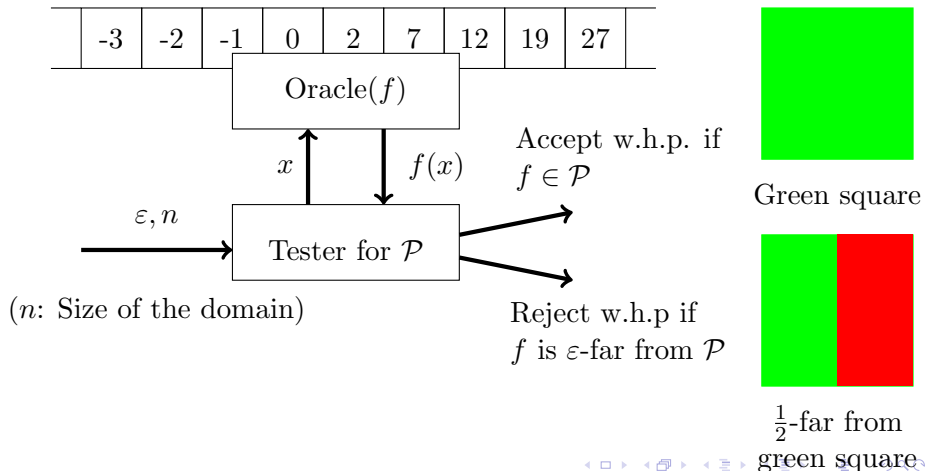
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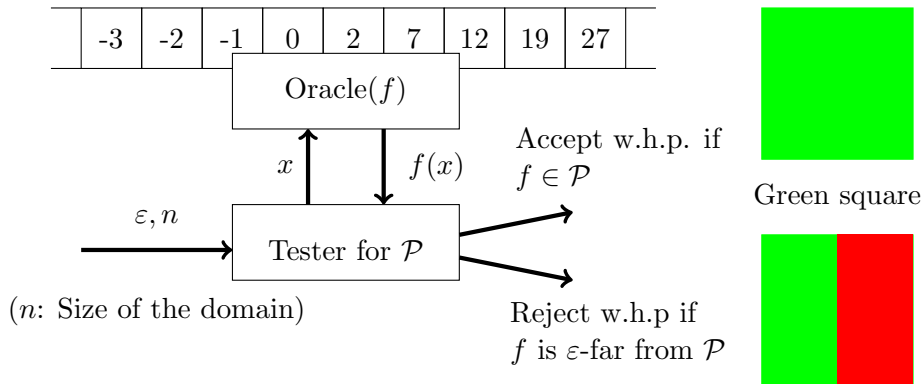
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Assumption: Oracle returns function values at all queried points.

Erasure-resilient property testing

[Dixit Raskhodnikova Thakurta Varma]

Decide w.p. $2/3$ if a function f , with $\leq \alpha$ fraction of values erased, satisfies a property \mathcal{P} or is ε -far from \mathcal{P} (w.r.t nonerased points).

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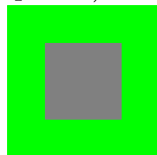
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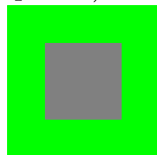


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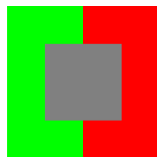
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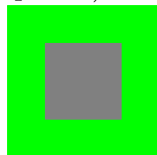
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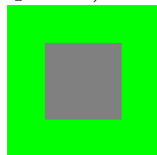
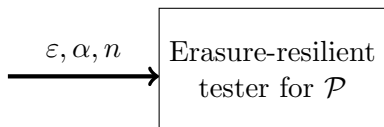
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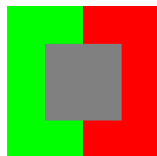
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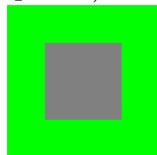
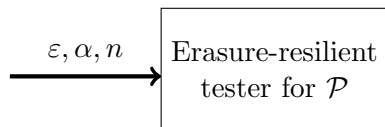
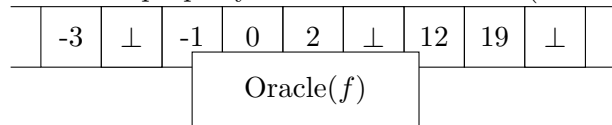


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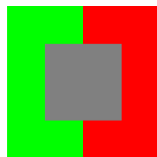
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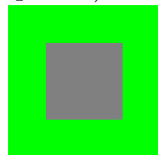
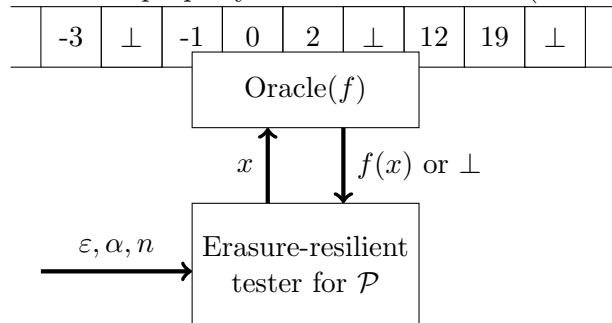


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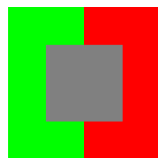
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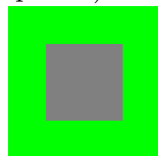
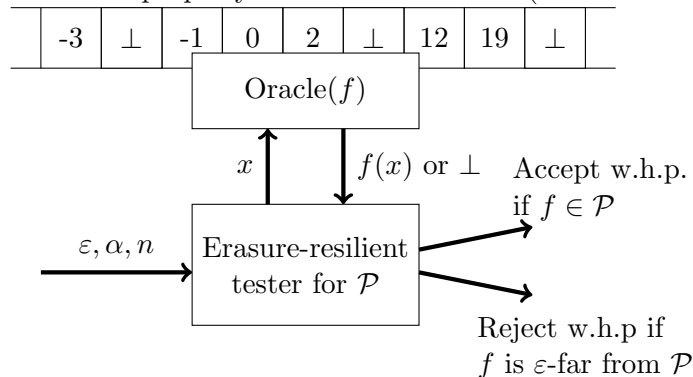


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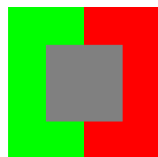
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Motivation: Some function values could be protected for privacy, or erased by an adversary.

Features of our model

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- Falls in between standard property testing and tolerant property testing [Parnas, Ron and Rubinfeld 06].
 - ▶ Standard testing : All function values are present and correct.
 - ▶ Tolerant testing : Some function values are adversarially corrupted.
 - ▶ Erasure-resilient testing : Some function values are adversarially erased.

Properties that we study

- 1 Monotonicity, Lipschitz property and convexity of real-valued functions.
- 2 Widely studied in property testing

[GGLRS00, DDGLRRS99, EKKR00, FLNRRS02, PRR03, F04, PRR06, BGJRW09, BCGM12, BBM12, CS13a, CS13b, JR13, CST14, BerRY14, BlaRY14, CDST15, KMS15, CDJS15, BB16, ..].

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- **convex** if $x < y < z \implies \frac{f(y) - f(x)}{y - x} \leq \frac{f(z) - f(y)}{z - y}$ for all $x, y, z \in [n]$.

Overview of our results

- Black box transformation of some ‘simple’ standard testers to efficient erasure-resilient testers.
- Efficient erasure-resilient testers in cases where our transformation does not apply.
- Existence of a property that has an ‘efficient’ standard tester and no ‘efficient’ erasure-resilient tester.

Our results: A black box transformation

- Makes **uniform testers for extendable properties** α -erasure-resilient.
- Uses the original testers as black box.

$O(\frac{1}{1-\alpha})$ factor query complexity overhead for all $\alpha \in [0, 1)$.

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- Applies to:
 - ▶ Monotonicity over general partial orders
[Fischer Lehman Newman Raskhodnikova Rubinfeld Samorodnitsky 02].
 - ▶ Convexity of black and white images
[Berman Murzabulatov Raskhodnikova 15].
 - ▶ Boolean functions having at most k alternations in values.

A limitation of our black box transformation

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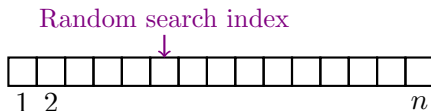
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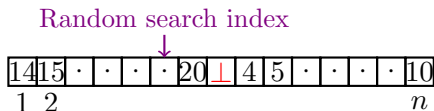


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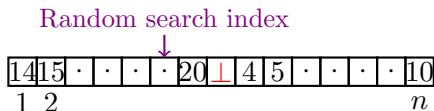


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Some of the known optimal testers for monotonicity, Lipschitz property and convexity also break on a constant number of erasures.

Our results

Erasure-resilient testers for monotonicity, Lipschitz property and convexity

For functions $f : [n] \rightarrow \mathbb{R}$

- α -erasure-resilient testers for monotonicity, Lipschitz property and convexity.

$O(\frac{1}{1-\alpha})$ factor query complexity overhead for all $\alpha \in [0, 1)$.

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For functions $f : [n]^d \rightarrow \mathbb{R}$

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- Hard example for our algorithms.
 - Our algorithms will not detect violations if $\alpha = \Omega(\frac{\epsilon}{\sqrt{d}})$.

Our results

Separation of erasure-resilient testing from standard testing

There exists a property \mathcal{P} and a constant $c > 0$ such that

- standard ε -testing of \mathcal{P} with $O\left(\frac{1}{\varepsilon}\right)$ queries
- α -erasure-resilient testing of \mathcal{P} needs $\Omega(n^c)$ queries \forall constant α .

Erasure-resilient testing is much harder than standard testing in general.

Summary of our contributions

- Definition of a model of property testing that accounts for erasures in the input data.
- A black box transformation from some simple property testers to erasure-resilient testers.
- Efficient erasure-resilient testers for monotonicity, Lipschitz property and convexity.
- A strong separation between testing in the presence of erasures and testing in the absence of erasures.

Erasure-resilient monotonicity tester

Theorem

There exists an α -erasure-resilient ε -tester for monotonicity of functions $f : [n] \rightarrow \mathbb{R}$ that makes $O\left(\frac{\log n}{\varepsilon(1-\alpha)}\right)$ queries for all $\alpha \in [0, 1), \varepsilon \in (0, 1)$.

Erasure-resilient monotonicity tester

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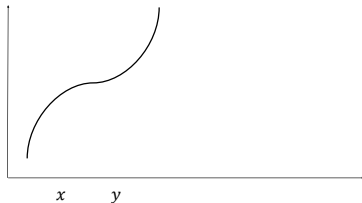
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- ε -testing of monotonicity using $\Theta\left(\frac{\log n}{\varepsilon}\right)$ queries [Ergün Kannan Kumar Rubinfeld Vishwanathan 00, Fischer 04].

Monotonicity

A partially erased function $f : [n] \rightarrow \mathbb{R}$ is monotone, if for all $x, y \in [n]$ that are nonerased

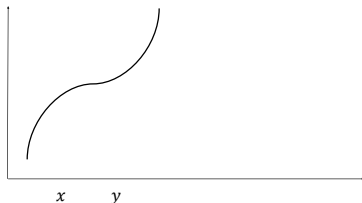
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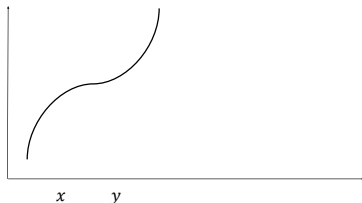


- **Violation:** Two nonerased points $x, y \in [n]$ such that $x < y$ and $f(x) > f(y)$.

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A partially erased function $f : [n] \rightarrow \mathbb{R}$ is monotone, if for all $x, y \in [n]$ that are nonerased

$$x < y \implies f(x) \leq f(y).$$

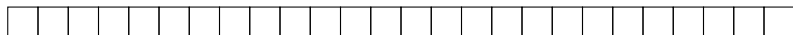


- **Violation:** Two nonerased points $x, y \in [n]$ such that $x < y$ and $f(x) > f(y)$.
- Monotonicity of functions over $[n] \equiv$ Sortedness of n -length arrays.

Our erasure-resilient sortedness tester

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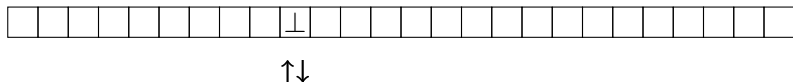
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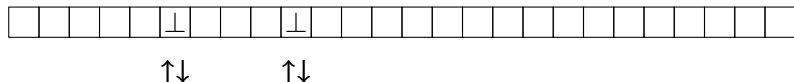
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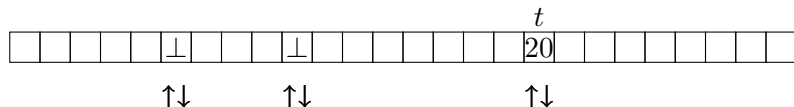
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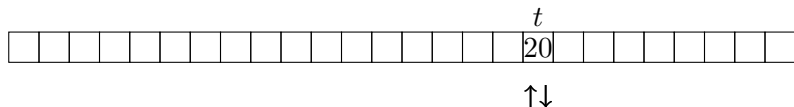
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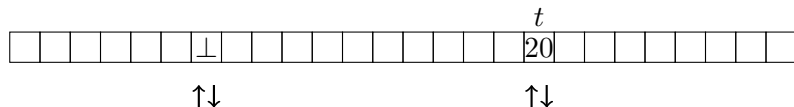
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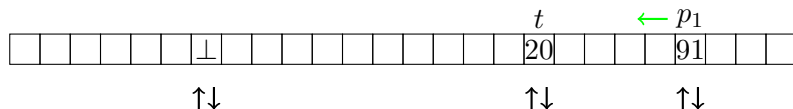
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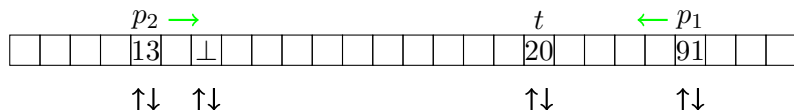
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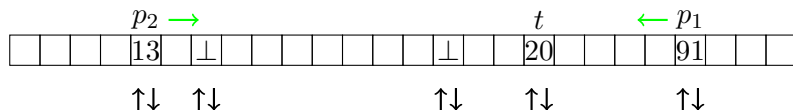
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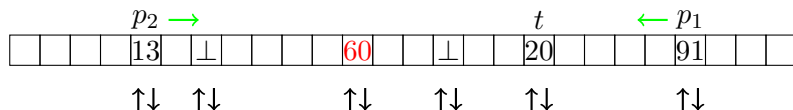
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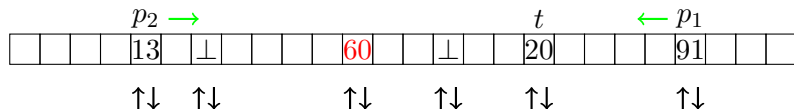
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 - Reject if there are violations along the search path.



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Query lemma

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- 1 Tester traverses a uniformly random search path in a random binary search tree.
- 2 The number of levels in a random binary search is $O(\log n)$ w.h.p.
- 3 Expected # of queries to one level of binary search is $O\left(\frac{1}{1-\alpha}\right)$.

Expected number of queries to a single level

Claim: $\mathbb{E}[\# \text{ of queries to one level of binary search}] = O\left(\frac{1}{1-\alpha}\right)$.

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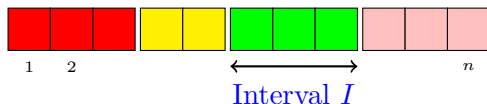
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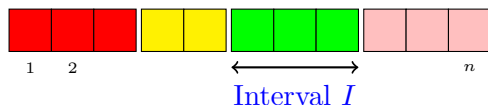


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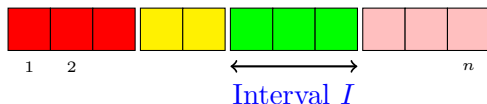
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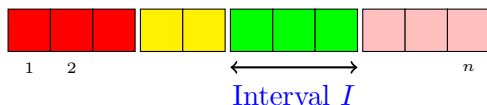
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$$\therefore \mathbb{E}[\# \text{ of queries to level } k] \leq \sum_{\text{intervals } I \text{ in level } k} \frac{|I|}{n(1-\alpha)} \leq \frac{1}{1-\alpha}.$$

Analysis overview

- ① Array is sorted \implies Tester accepts.
- ② Array is ε -far from sorted \implies Detects violation w.p. $\geq \varepsilon$ in an iteration.
 - ▶ In $2/\varepsilon$ independent iterations:

$$\Pr[\text{tester does not detect a violation}] \leq (1 - \varepsilon)^{2/\varepsilon} \leq \frac{1}{3}.$$

Tester rejects with high constant probability.

- ③ The expected number of queries made is $O\left(\frac{\log n}{\varepsilon(1-\alpha)}\right)$.

What we showed

Theorem

There exists an α -erasure-resilient ε -tester for **sortedness** of n -length arrays that makes $O\left(\frac{\log n}{\varepsilon(1-\alpha)}\right)$ queries for all $\alpha \in [0, 1), \varepsilon \in (0, 1)$.

Ongoing work

- Separating erasures from corruptions.
 - ▶ Is tolerant testing harder than erasure-resilient testing in general?
- Adversarial erasures vs. random erasures

Future directions

- Better erasure-resilient testers for monotonicity and Lipschitz property of functions $f : [n]^d \mapsto \mathbb{R}$.
- Other models of erasures
 - ▶ Random erasures.
 - ▶ Semi-adaptive erasures.

Thank you!