

Does Locality imply Efficient Testability?

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WOLA 2019

Monotonicity testing: Yet another proof..

Consider an array of numbers. Is the array monotone increasing?



array is monotone



array not monotone

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Property Testing:

Given query access to $A: [n] \rightarrow \mathbb{R}$ that is ϵ -far from being monotone increasing, how many queries needed to find (with prob. $2/3$) a “proof” that A is not monotone.

ϵ -far:

Need to change ϵn entries in A to make it monotone.

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[Ergün, Kannan, Kumar, Rubinfeld, Viswanthan '98:]

Monotonicity is ε -testable with $O(\varepsilon^{-1} \log n)$ queries.

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Monotonicity testing: Yet another proof..

1	13	17	19	2	3	5	7	20	22	21	31	41	47	60	59
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Consider a *partitioning* of the array into intervals.

In which intervals is the *first elements larger than the last?*

Monotonicity testing: Yet another proof..



Hierarchical partitioning:

each interval in level i is
union of two or three
intervals from level $i - 1$

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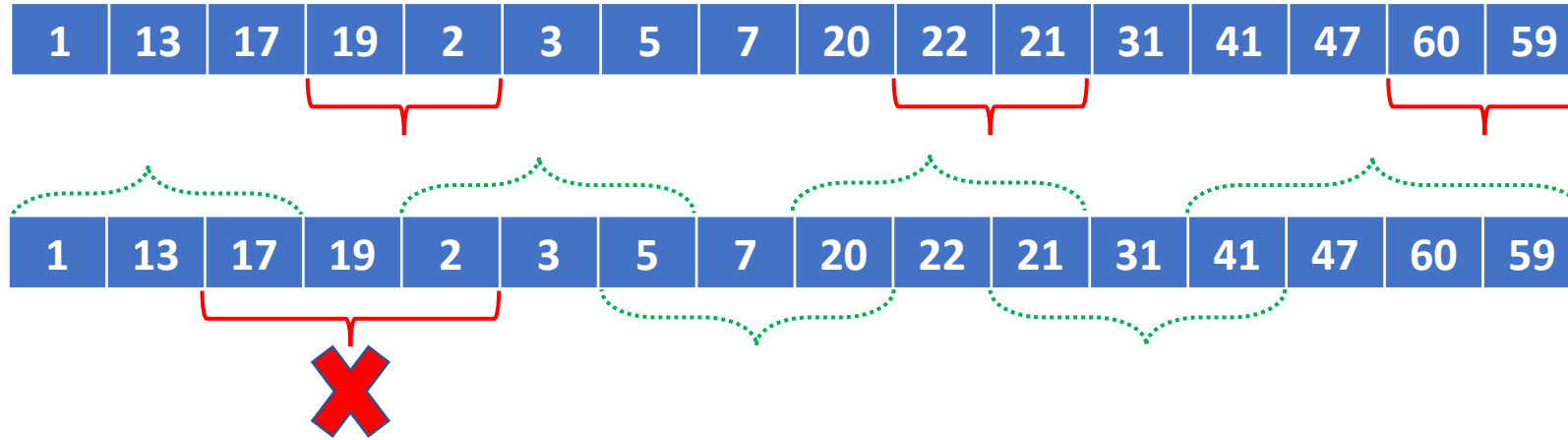
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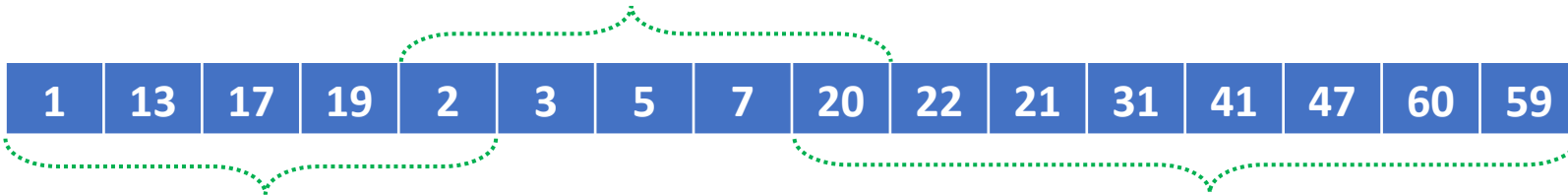
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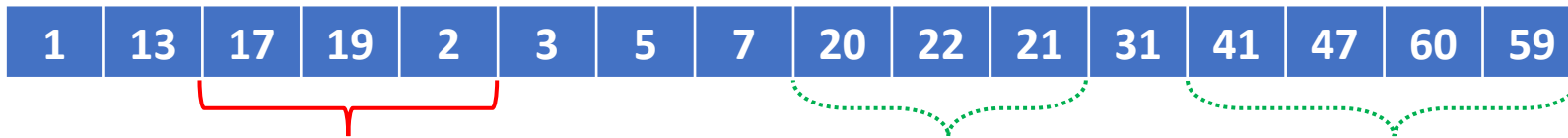
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Monotonicity testing: Yet another proof..



Consider only "bad" intervals that are maximal: not contained in any other "bad" one.



Claim: Suffices to edit elements within "good" intervals that are one level above maximal "bad" ones, to make array monotone

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1	13	17	19	2	3	5	7	20	$20\frac{1}{2}$	21	31	41	47	53	59
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Corollary: If array is ε -far from monotonicity, then set of "maximal **bad** intervals" has size st least $\approx \varepsilon n$

The test: Pick $\approx 1/\varepsilon$ intervals from each level, query their endpoints. Reject if any of them is **bad**. Total query complexity $\approx \varepsilon^{-1} \log n$.

Local properties

A property of arrays $A: [n] \rightarrow \Sigma$ is **k -local** if it can be defined by a family of **forbidden consecutive patterns** of size $\leq k$.

Examples:

Monotonicity is 2-local. Forbidden patterns: " $A(i) > A(i + 1)$ "



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Convexity is **3-local**

Properties of first k discrete derivatives are **$(k + 1)$ -local**

Pattern matching and computational biology problems are **k -local** for small k

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Submodularity is **2-local**

Properties of first k discrete derivatives are **$(k + 1)$ -local**

Pattern matching problems in computer vision are **k -local** for small k

Local properties \Leftrightarrow Local algorithms

The LOCAL model in distributed computing [Linial'87]:

Which graph properties are “locally decidable” by balls of radius k ?

Our setting a bit different:

1. **Graph topology known in advance:** graph is the line (for $d=1$) / hypergrid ($d>1$).
2. However, each **vertex** holds a **value (not known in advance)**.

Claim: Property is k -local \Leftrightarrow has local algorithm (known topology, unknown values) with $\Theta(k)$ rounds

Generic test for local properties

Theorem [B., 2019]:

Any k -local property \mathcal{P} of $[n]^d$ -arrays over any finite alphabet Σ is ε -testable using

$$O\left(\frac{k \log n}{\varepsilon}\right) \text{ queries for } d = 1$$

$$O_d\left(\frac{kn^{d-1}}{\varepsilon^{1/d}}\right) \text{ queries for } d > 1$$

Property Testing:

Given property \mathcal{P} , parameter ε , and query access to $A: [n]^d \rightarrow \Sigma$, distinguish with prob. $2/3$ between the cases:

- A satisfies \mathcal{P}
- A is ε -far from \mathcal{P} : need to change εn^d values in A to satisfy \mathcal{P}

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The good news: Test is **canonical** (queries depend on d, k, ε, n , but not on \mathcal{P}, Σ);
proximity oblivious (repetitive iterations of the same “basic” test);
non-adaptive (makes all queries in advance); and has
one-sided error.

Allows “**sketching for testing**”.

The bad news: linear running time for $d = 1$; exponential for $d > 1$ ☹

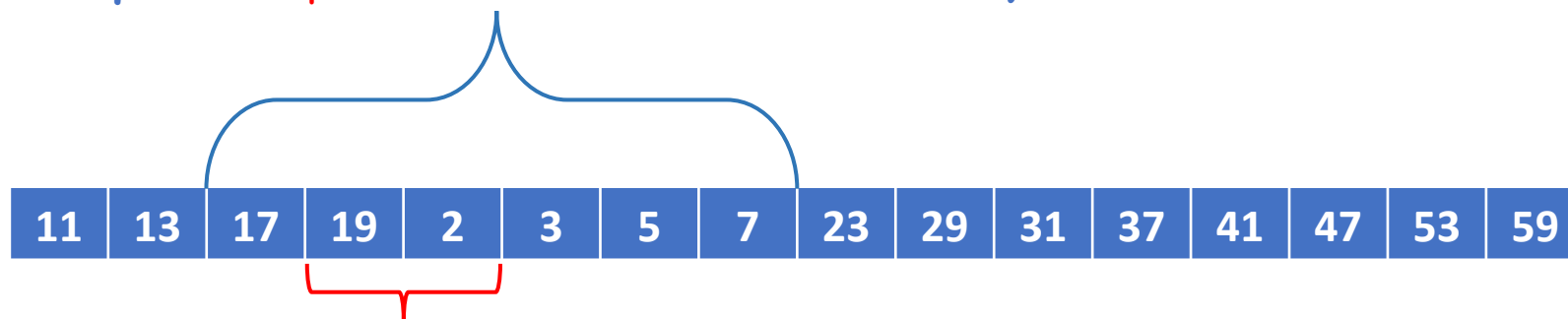
The main idea: Unrepairability

\mathcal{P} : a 2-local property of 1D arrays $A: [n] \rightarrow \Sigma$.

An interval $I = \{a, a + 1, \dots, b\} \subseteq [n]$ is **unrepairable** (w.r.t A, \mathcal{P}) if, no matter how we modify $A(a + 1), \dots, A(b - 1)$, the sub-array of A between a and b will **never** satisfy \mathcal{P} .

Observation: Enough to query only $f(a)$ and $f(b)$ to know if I is unrepairable.

Example: **unrepairable** interval for **monotonicity**.



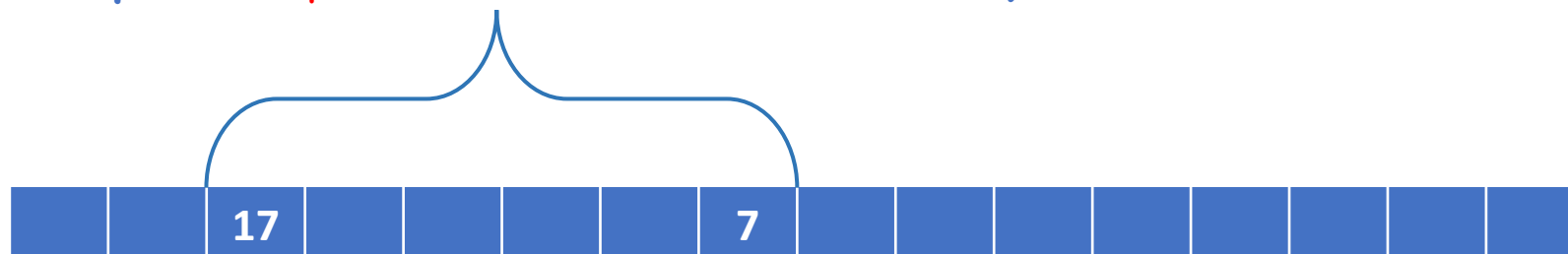
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The main idea: Unrepairability

\mathcal{P} : a 2-local property of 1D arrays $A: [n] \rightarrow \Sigma$.

Proof idea:

Structural result: Suppose that A is ε -far from \mathcal{P} . Then there is a set of “canonical” **unrepairable** intervals covering $\geq \varepsilon n$ of the entries.

Algorithm: For any $i = 0, 1, \dots, \log n$, pick $\approx 1/\varepsilon$ “canonical” intervals of length $\approx 2^i$ and query their endpoints.

With good probability, one of the intervals will be unrepairable.

Extension to multiple dimensions:

Replace “intervals” by “ d -dimensional consecutive boxes” and “endpoints” with “ $(d - 1)$ -dimensional boundaries”.

1	4	3	5				
2			7				
8			9				
3	4	4	6				

Non-adaptive Lower bounds

The upper bound is tight for non-adaptive algorithms, for any fixed $d \geq 1$

For $d = 1$, matches $\Theta(\log n)$ bounds for monotonicity [EKKRV'98, F'04, CS'13], convexity [PRR'04], and Lipschitz [JR'11]. Tight for monotonicity even among adaptive two-sided tests.

For $d > 1$,

Theorem [B., 2019]:

There exists a k -local property of $[n]^d$ -arrays over alphabet of size $n^{O(d)}$, whose non-adaptive one-sided query complexity is $\Omega_d(k\varepsilon^{-\frac{1}{d}}n^{d-1})$.

Non-adaptive Lower bounds

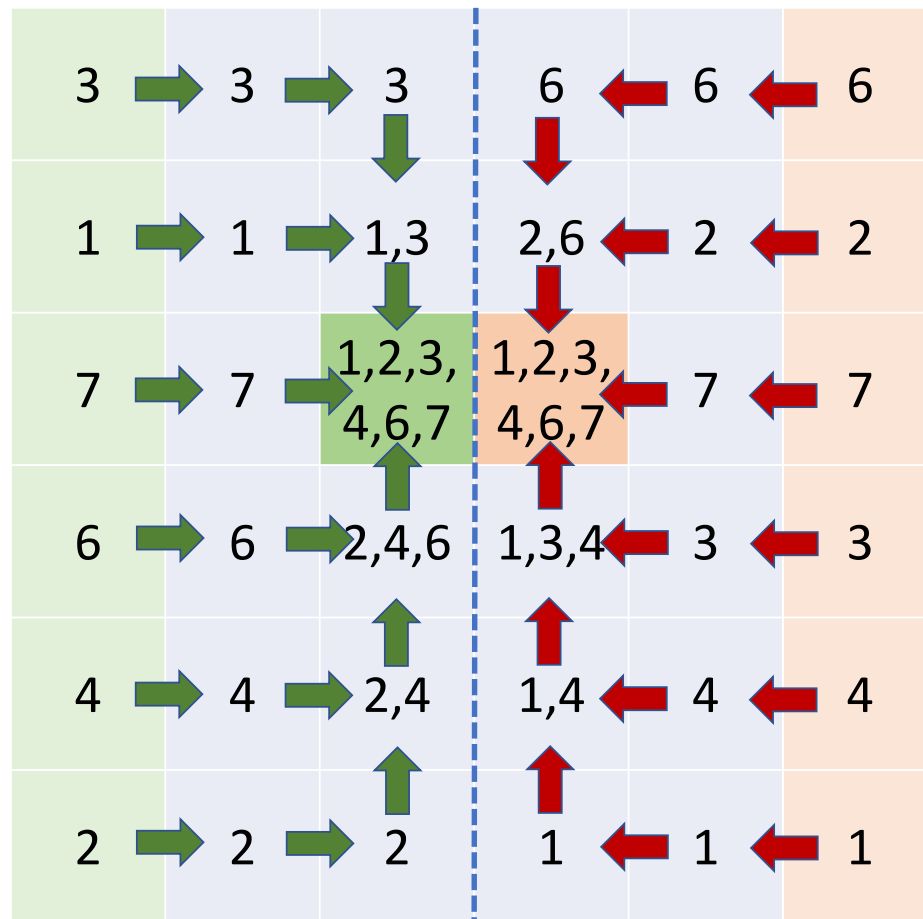
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Adaptive Lower bounds

What about the adaptive case for $d > 1$?

Theorem [B., 2019+]:

There exists a 2-local property of $[n]^d$ -arrays, whose adaptive two-sided query complexity is $n^{\Omega(1)}$.

Open question: close the gaps – no known lower bounds depending on d .

Questions

1. Exponential running time is undesirable.
[S. Raskhodnikova, C. Seshadhri:] For which subclasses of local properties can we also get sublinear running time?
[Chakrabarty, Seshadhri '12]: “bounded derivative” properties.
2. On which graph does “locality \Rightarrow sublinear testability” hold?
Bounded-degree graphs? Hyperfinite graphs?
3. How powerful is adaptivity?

Thank you!