Finding monotone patterns

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ICALP

July 2022

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- or is far from having property \mathcal{P} .

We consider testing with **one-sided error**: given an object which is far from having \mathcal{P} , provide evidence of not being in \mathcal{P} , with high probability.

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We sometimes refer to an increasing k-tuple as a (1...k)-copy.

History



Ergün–Kannan–Kumar–Rubinfeld–Viswanathan '98. Optimal non-adaptive monotonicity testers make $\Theta(\log n)$ queries.

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Newman–Rabinovich–Rajendraprasad–Sohler '17. For $k \ge 2$, there is a (non-adaptive) tester which makes (log n)^{$O(k^2)$} queries.

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An optimal non-adaptive algorithm for testing (1...k)-freeness makes $\Theta_k((\log n)^{\lfloor \log_2 k \rfloor})$ queries.

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An optimal adaptive algorithm for testing (1...k)-freeness makes $\Theta_k(\log n)$ queries.

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- $\Theta(1)$ queries in $[\ell, \ell + 2^i]$ for $i \in [\log n]$ and sampled ℓ .

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Lemma

Let \mathcal{I} be a family of disjoint intervals in [n] s.t. $\sum_{I \in \mathcal{I}} |I| \ge \alpha n$.



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Thus, if *f* is structured, there is a **robust collection** of disjoint 'splittable intervals' \mathcal{J} s.t.

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$$\sum_{J\in\mathcal{J}}|J|=\Omega(n),$$

• if K contains $J \in \mathcal{J}$ then K has $\Omega(|K|)$ disjoint increasing k-tuples.

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- If y^{*} ≈ y: Θ(log n) queries to find increasing s-tuple π₁ near x and (k − s)-tuple π₂ near y^{*} and above π₁.



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1 J_i has $\Omega(|J_i|)$ increasing (i + 1)-tuples strictly below $f(y^*)$ (case Ai), or **2** J_i has $\Omega(|J_i|)$ increasing (k - i)-tuples above $f(y^*)$ (case Bi).



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