#### Improved Sublinear-Time Algorithms for Testing Order Pattern Freeness

Ilan Newman





Nithin Varma





100 98 723 1.2 5.68 3 1

• Let *A* be an array of length *n* and  $\pi : [k] \rightarrow [k]$  be a bijection

100 98 723 1.2 5.68 3 1

- Let *A* be an array of length *n* and  $\pi : [k] \rightarrow [k]$  be a bijection
- Array *A* has a <u> $\pi$ -appearance</u> if  $\exists$  indices  $i_1 < \ldots < i_k$  such that  $A[i_a] > A[i_b]$  if  $\pi(a) > \pi(b) \quad \forall a, b \in [k]$

100 98 723 1.2 5.68 3 1

- Let *A* be an array of length *n* and  $\pi : [k] \rightarrow [k]$  be a bijection
- Array *A* has a <u> $\pi$ -appearance</u> if  $\exists$  indices  $i_1 < \ldots < i_k$  such that  $A[i_a] > A[i_b]$  if  $\pi(a) > \pi(b) \quad \forall a, b \in [k]$
- A is  $\pi$ -free if it has no  $\pi$ -appearance

 100
 98
 723
 1.2
 5.68
 3
 1

- Let *A* be an array of length *n* and  $\pi : [k] \rightarrow [k]$  be a bijection
- Array *A* has a <u>*π*-appearance</u> if ∃ indices  $i_1 < ... < i_k$  such that  $A[i_a] > A[i_b]$  if  $\pi(a) > \pi(b) \quad \forall a, b \in [k]$
- A is  $\pi$ -free if it has no  $\pi$ -appearance
- Above array has a (3,4,1,2)-appearance but is (1,2,3,4)-free

• Well studied notion in combinatorics [Bona; J. Comb. Theory '99], [Arratia; Elec. J. Of Comb. '99], [Alon & Friedgut; J. Comb. Theory '00], [Marcus & Tardos; J. Comb. Theory '04]

- Well studied notion in combinatorics [Bona; J. Comb. Theory '99], [Arratia; Elec. J. Of Comb. '99], [Alon & Friedgut; J. Comb. Theory '00], [Marcus & Tardos; J. Comb. Theory '04]
- Several classical algorithms to decide π-freeness [Albert, Aldred, Atkinson & Holton; ISAAC '01], [Ahal & Rabinovich; SIDMA '08], [Guillemot & Marx; SODA '14], [Berendson, Kozma & Marx; Algorithmica '21]

- Well studied notion in combinatorics [Bona; J. Comb. Theory '99], [Arratia; Elec. J. Of Comb. '99], [Alon & Friedgut; J. Comb. Theory '00], [Marcus & Tardos; J. Comb. Theory '04]
- Several classical algorithms to decide π-freeness [Albert, Aldred, Atkinson & Holton; ISAAC '01], [Ahal & Rabinovich; SIDMA '08], [Guillemot & Marx; SODA '14], [Berendson, Kozma & Marx; Algorithmica '21]
- Motivated by detecting motifs and patterns in time series analysis [Berndt & Clifford; AAAI '94], [Keogh, Lonardi & Chiu; SIGKDD '02]

- Well studied notion in combinatorics [Bona; J. Comb. Theory '99], [Arratia; Elec. J. Of Comb. '99], [Alon & Friedgut; J. Comb. Theory '00], [Marcus & Tardos; J. Comb. Theory '04]
- Several classical algorithms to decide  $\pi$ -freeness[Albert, Aldred, Atkinson & Holton; ISAAC '01], [Ahal & Rabinovich; SIDMA '08], [Guillemot & Marx; SODA '14], in particular a linear time algorithm [Berendson, Kozma & Marx; Algorithmica '21]
- Motivated by detecting motifs and patterns in time series analysis [Berndt & Clifford; AAAI '94], [Keogh, Lonardi & Chiu; SIGKDD '02]
- Connections to Longest Increasing Subsequence [Newman & V.; ICALP '21]

- Well studied notion in combinatorics [Bona; J. Comb. Theory '99], [Arratia; Elec. J. Of Comb. '99], [Alon & Friedgut; J. Comb. Theory '00], [Marcus & Tardos; J. Comb. Theory '04]
- Several classical algorithms to decide  $\pi$ -freeness[Albert, Aldred, Atkinson & Holton; ISAAC '01], [Ahal & Rabinovich; SIDMA '08], [Guillemot & Marx; SODA '14], in particular a linear time algorithm [Berendson, Kozma & Marx; Algorithmica '21]
- Motivated by detecting motifs and patterns in time series analysis [Berndt & Clifford; AAAI '94], [Keogh, Lonardi & Chiu; SIGKDD '02]
- Connections to Longest Increasing Subsequence [Newman & V.; ICALP '21]
   Today: Sublinear-Time Algorithms for (approximate) Pattern Freeness

#### **Decision Problem**



#### **Decision Problem**

• Cannot exactly solve nontrivial decision problems without full access to the input



Property testing

[Rubinfeld & Sudan; **SODA 92 & SICOMP 96**, Goldreich, Goldwasser & Ron; **FOCS 96 & JACM 98**]

• <u>*ɛ*-far from property</u>: At least *ɛ* fraction of input values need to be changed to satisfy the property



#### Property testing

[Rubinfeld & Sudan; **SODA 92 & SICOMP 96**, Goldreich, Goldwasser & Ron; **FOCS 96 & JACM 98**]



Property testing

[Rubinfeld & Sudan; **SODA 92 & SICOMP 96**, Goldreich, Goldwasser & Ron; **FOCS 96 & JACM 98**]

- Several well-studied and fundamental properties
  - Bipartiteness [Alon, Krivelevich; SICOMP '02, ...]
  - Monotonicity [Chakrabarty Seshadhri; STOC '13, Khot, Minzer, Safra; SICOMP '18, ...]
  - Convexity of images [Berman Murzabulatov Raskhodnikova; RSA '19, ...]
  - Linearity and related properties [Blum Luby Rubinfeld; JCSS '93, ...]



- Given query access to an array *A* of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether
  - A is  $\pi$ -free, OR
  - *A* is  $\epsilon$ -far from  $\pi$ -free

• Given query access to an array A of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether



• Given query access to an array A of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether



Generalization of monotonicity testing of arrays [EKKRV00, DGLRRS99, F04, ...]

• Given query access to an array A of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether



```
Monotonicity \equiv (2,1)-freeness
```

Generalization of monotonicity testing of arrays [EKKRV00, DGLRRS99, F04, ... ]

- Given query access to an array A of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether
  - A is  $\pi$ -free, OR • A is  $\epsilon$ -far from  $\pi$ -free • A is  $\epsilon$ -far from  $\pi$ -free

Algorithm solving this problem is called  $\underline{\epsilon}$ -tester for  $\pi$ -freeness

- Given query access to an array A of length *n*, a parameter  $\epsilon \in (0,1)$ , and permutation  $\pi$ , decide whether
  - A is  $\pi$ -free, OR • A is  $\epsilon$ -far from  $\pi$ -free • A is  $\epsilon$ -far from  $\pi$ -free

Algorithm solving this problem is called  $\underline{\epsilon}$ -tester for  $\pi$ -freeness

Tester <u>adaptive</u> if its queries depend on answers to previous queries

Tester <u>nonadaptive</u> otherwise

• **Special case**: Sortedness testing of arrays of length *n* [Ergun Kannan Kumar Rubinfeld Vishwanathan 00, Fischer 04, ... ]

- **Special case**: Sortedness testing of arrays of length *n* [Ergun Kannan Kumar Rubinfeld Vishwanathan 00, Fischer 04, ... ]
- [Newman Rabinovich Rajendraprasad Sohler 19] initiated the study of sublinear-time algorithms for testing pattern freeness for larger patterns

- **Special case**: Sortedness testing of arrays of length *n* [Ergun Kannan Kumar Rubinfeld Vishwanathan 00, Fischer 04, ... ]
- [Newman Rabinovich Rajendraprasad Sohler 19] initiated the study of sublinear-time algorithms for testing pattern freeness for larger patterns
- Studied in depth for monotone patterns (1, 2, ..., k) or (k, k 1, ..., 1)

[NRRS19, BenEliezer Canonne Letzter Waingarten 19, BenEliezer Letzter Waingarten 19]

• Sortedness testing of arrays of length n can be done using  $\Theta(\log n)$  queries [EKKRV00, F04]

- Sortedness testing of arrays of length n can be done using  $\Theta(\log n)$  queries [EKKRV00, F04]
- *O*(log *n*)-query tester for monotone patterns of constant length [BLW19]
- polylog *n* query tester for arbitrary patterns of length 3 [NRRS19]

- Sortedness testing of arrays of length n can be done using  $\Theta(\log n)$  queries [EKKRV00, F04]
- *O*(log *n*)-query tester for monotone patterns of constant length [BLW19]
- polylog n query tester for arbitrary patterns of length 3 [NRRS19]

What about nonmonotone patterns of length > 3?

- Sortedness testing of arrays of length n can be done using  $\Theta(\log n)$  queries [EKKRV00, F04]
- *O*(log *n*)-query tester for monotone patterns of constant length [BLW19]
- polylog *n* query tester for arbitrary patterns of length 3 [NRRS19]

What about nonmonotone patterns of length > 3?

- Nonadaptive testers making  $O(n^{1-\frac{1}{k-1}})$  queries [NRRS19]
- Nonadaptive testers cannot do better! [NRRS19, BenEliezer Canonne 18]

- Sortedness testing of arrays of length n can be done using  $\Theta(\log n)$  queries [EKKRV00, F04]
- *O*(log *n*)-query tester for monotone patterns of constant length [BLW19]
- polylog *n* query tester for arbitrary patterns of length 3 [NRRS19]

What about nonmonotone patterns of length > 3? • Nonadaptive testers making  $O(n^{1-\frac{1}{k-1}})$  queries [NRRS19]

• Nonadaptive testers cannot do better! [NRRS19, BenEliezer Canonne 18] What about adaptive testers ?

#### Our Result

Let  $\epsilon \in (0,1), k \in \mathbb{N}, \pi \in \mathbb{S}_k$ . There is an  $\epsilon$ -tester for  $\pi$ -freeness

- with query complexity  $\tilde{O}(n^{o(1)})$
- that always accepts  $\pi$ -free arrays.

#### Our Result

Let  $\epsilon \in (0,1), k \in \mathbb{N}, \pi \in \mathbb{S}_k$ . There is an  $\epsilon$ -tester for  $\pi$ -freeness

- with query complexity  $\tilde{O}(n^{o(1)})$
- that always accepts  $\pi$ -free arrays.
- Improved sublinear-time guarantee
- Our techniques are general and do not rely on the structure of the pattern being considered (e.g., as opposed to [NRRS19])

## Today

•  $\tilde{O}(\sqrt{n})$ -query algorithm to test  $\pi$ -freeness of  $\pi \in \mathbb{S}_4$ 

## Today

- $\tilde{O}(\sqrt{n})$ -query algorithm to test  $\pi$ -freeness of  $\pi \in \mathbb{S}_4$
- [BC18] Every nonadaptive algorithm has query complexity  $\Omega(n^{2/3})$

## Today

- $\tilde{O}(\sqrt{n})$ -query algorithm to test  $\pi$ -freeness of  $\pi \in \mathbb{S}_4$
- [BC18] Every nonadaptive algorithm has query complexity  $\Omega(n^{2/3})$

**Goal**: Design an algorithm that makes  $\tilde{O}(\sqrt{n})$  queries and finds a  $\pi$ -appearance in an array that is  $\epsilon$ -far from  $\pi$ -free

• Array A is  $\epsilon$ -far from being  $\pi$ -free  $\implies$  there are  $\geq \frac{\epsilon n}{4}$  disjoint  $\pi$ -appearances

• Array A is  $\epsilon$ -far from being  $\pi$ -free  $\implies$  there are  $\geq \frac{\epsilon n}{4}$  disjoint  $\pi$ -appearances

• Used by all algorithms to test  $\pi$ -freeness

• Array A is  $\epsilon$ -far from being  $\pi$ -free  $\implies$  there are  $\geq \frac{\epsilon n}{4}$  disjoint  $\pi$ -appearances

• Used by all algorithms to test  $\pi$ -freeness

• Array A is  $\epsilon$ -far from being  $\pi$ -free  $\implies$  there are  $\geq \frac{\epsilon n}{4}$  disjoint  $\pi$ -appearances

• Used by all algorithms to test  $\pi$ -freeness

Disjoint (4,3,2,1)-appearances

• (Conceptual) Step 1: View the array as a box of *n* points in  $[n] \times \mathbb{R}$ 



• (Conceptual) Step 1: View the array as a box of *n* points in  $[n] \times \mathbb{R}$ 





- (Conceptual) Step 1: View the array as a box of *n* points in  $[n] \times \mathbb{R}$
- Step 2: Query  $\tilde{O}(\sqrt{n})$  uniform indices to approximate the box by a coarse  $O(\sqrt{n}) \times \sqrt{n}$  grid of sub-boxes



- (Conceptual) Step 1: View the array as a box of *n* points in [*n*] × ℝ
- Step 2: Query  $\tilde{O}(\sqrt{n})$  uniform indices to approximate the box by a coarse  $O(\sqrt{n}) \times \sqrt{n}$  grid of sub-boxes

We ensure that rows and columns roughly equipartition the set of array points



- (Conceptual) Step 1: View the array as a box of *n* points in [*n*] × ℝ
- Step 2: Query  $\tilde{O}(\sqrt{n})$  uniform indices to approximate the box by a coarse  $O(\sqrt{n}) \times \sqrt{n}$  grid of sub-boxes

Each row has  $\sim \sqrt{n}$  points Each column has  $\sqrt{n}$  points



### Gridding: Part 2

- From each column, sample  $\tilde{O}(1)$  uniformly random points
- Tag a box as **nonempty** if it has at least one sampled point; tag it **dense** if it has a constant fraction of sampled points
- There are O(n) boxes, out of which we only tag  $\tilde{O}(\sqrt{n})$  boxes



## Gridding: Part 2

- From each column, sample  $\tilde{O}(1)$  uniformly random points
- Tag a box as **nonempty** if it has at least one sampled point; tag it **dense** if it has a constant fraction of sampled points
- There are O(n) boxes, out of which we only tag  $\tilde{O}(\sqrt{n})$  boxes

If  $\pi = (3,2,1,4)$ , we already found a  $\pi$ -appearance



#### In general, if too many nonempty boxes...

• [Markus & Tardos '04] For each permutation  $\pi$ , there exists a constant  $\kappa$ such that such that if the grid on the right has more than  $\kappa \sqrt{n}$  nonempty boxes, then there is a  $\pi$ -appearance among the nonempty boxes



#### We show that:

• Only  $O(\sqrt{n})$  nonempty boxes. In particular, only  $O(\sqrt{n})$  dense boxes



#### We show that:

- Only  $O(\sqrt{n})$  nonempty boxes. In particular, only  $O(\sqrt{n})$  dense boxes
- Each row and column contains O(1) dense boxes each



#### We show that:

- Only  $O(\sqrt{n})$  nonempty boxes. In particular, only  $O(\sqrt{n})$  dense boxes
- Each row and column contains O(1) dense boxes each
- The union of points in all the dense boxes contains  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances



#### We show that:

- Only  $O(\sqrt{n})$  nonempty boxes. In particular, only  $O(\sqrt{n})$  dense boxes
- Each row and column contains O(1) dense boxes each
- The union of points in all the dense boxes contains  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances

**Ideas in Proof**: Combinatorial lemma by [MarcusTardos04] + Averaging arguments + Properties of the sampling procedure



#### We show that:

- Only  $O(\sqrt{n})$  nonempty boxes. In particular, only  $O(\sqrt{n})$  dense boxes
- Each row and column contains O(1) dense boxes each
- The union of points in all the dense boxes contains  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances

Where in the grid do these appear?









#### How to find $\pi$ -appearances in these configurations?

- For each configuration  $\mathscr{C}$ :
  - Check for  $\pi$ -appearances whose legs are in boxes forming the configuration  $\mathscr{C}$

• Suppose  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances have their legs distributed according to the following 2-boxed configuration 3



- Suppose  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances have their legs distributed according to the following 2-boxed configuration 3
- At most  $\sqrt{n}$  dense boxes & each dense box has  $\leq \sqrt{n}$  points



- Suppose  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances have their legs distributed according to the following 2-boxed configuration
- At most  $\sqrt{n}$  dense boxes & each dense box has  $\leq \sqrt{n}$  points



• A uniformly random dense box *B* contains the (3,2) legs of  $\Omega(\epsilon \sqrt{n})$  disjoint  $\pi$  -appearances

- Suppose  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances have their legs distributed according to the following 2-boxed configuration
- At most  $\sqrt{n}$  dense boxes & each dense box has  $\leq \sqrt{n}$  points



- A uniformly random dense box *B* contains the (3,2) legs of  $\Omega(\epsilon \sqrt{n})$  disjoint  $\pi$  -appearances
- There are O(1) dense boxes sharing a row with B, by our Gridding

- Suppose  $\Omega(\epsilon n)$  disjoint  $\pi$ -appearances have their legs distributed according to the following 2-boxed configuration
- At most  $\sqrt{n}$  dense boxes & each dense box has  $\leq \sqrt{n}$  points



- A uniformly random dense box *B* contains the (3,2) legs of  $\Omega(\epsilon \sqrt{n})$  disjoint  $\pi$  -appearances
- There are O(1) dense boxes sharing a row with B, by our Gridding
- Querying all points in the subarrays corresponding to *B* and the other dense boxes in its row is sufficient to detect such a  $\pi$ -appearance

 $\tilde{O}(\sqrt{n})$  queries overall, since we want a high success probability

- Most other configurations are dealt with by similar averaging arguments
- However, more complicated arguments needed for some specific configurations

- Most other configurations are dealt with by similar averaging arguments
- However, more complicated arguments needed for some specific configurations



- Most other configurations are dealt with by similar averaging arguments
- However, more complicated arguments needed for some specific configurations
- Our algorithm works for patterns of all constant length *k* 
  - For patterns of length k > 4, our recursive algorithm ends up solving a more general problem
  - Detect a  $\nu$ -appearance with a specific leg distribution, where  $\nu$  is some subpattern of  $\pi$



- Most other configurations are dealt with by similar averaging arguments
- However, more complicated arguments needed for some specific configurations
- Our algorithm works for patterns of all constant length *k* 
  - For patterns of length k > 4, our recursive algorithm ends up solving a more general problem
  - Detect a  $\nu$ -appearance with a specific leg distribution, where  $\nu$  is some subpattern of  $\pi$
- The algorithm is called recursively in order to improve the query complexity to  $\tilde{O}(n^{o(1)})$  from  $O(\sqrt{n})$

• What is the true complexity of testing  $\pi$ -freeness?

- What is the true complexity of testing  $\pi$ -freeness?
  - Lower bounds for adaptive algorithms?

- What is the true complexity of testing  $\pi$ -freeness?
  - Lower bounds for adaptive algorithms?
- What about patterns of superconstant length?

- What is the true complexity of testing  $\pi$ -freeness?
  - Lower bounds for adaptive algorithms?
- What about patterns of superconstant length?
- Approximating the distance of arrays to  $\pi$ -freeness

- What is the true complexity of testing  $\pi$ -freeness?
  - Lower bounds for adaptive algorithms?
- What about patterns of superconstant length?
- Approximating the distance of arrays to  $\pi$ -freeness

Thank you!