

# Lecture 18: Uniformity Testing Monotonicity Testing



# Administrivia, Plan

- Admin:
  - PS3: pick up
  - Project proposals: Nov 16th
  
- Plan:
  - Uniformity Testing
  - Monotonicity Testing
  
- Scriber?

# Uniformity testing

- Enough to distinguish:
  - $\|D\|_2^2 = 1/n$  (unif)
  - $\|D\|_2^2 > 1/n + \epsilon^2/n$  (non-unif)
- **Lemma:**  $\frac{1}{M} \cdot C$  is a good enough as long as
  - $m = \Omega\left(\frac{\sqrt{n}}{\epsilon^4}\right)$
  - where  $M = m(m-1)/2$
- Let  $d = \|D\|_2^2$
- Claim 1:  $E\left[\frac{C}{M}\right] = d$
- Claim 2:  $Var\left[\frac{C}{M}\right] \leq \frac{d}{M} + \frac{8d^2\sqrt{n}}{m}$
- Finish lemma proof...

Algorithm UNIFORM:

Input:  $n, m, x_1, \dots, x_m$

$C = 0;$

for( $i=0; i<m; i++$ )

  for( $j=i+1; j<m; j++$ )

    if ( $x_i = x_j$ )

$C++;$

if ( $C < a \cdot m^2/n$ )

  return “Uniform”;

else

  return “Not uniform”;

//  $a$ : constant dependent on  $\epsilon$



# Identity Testing

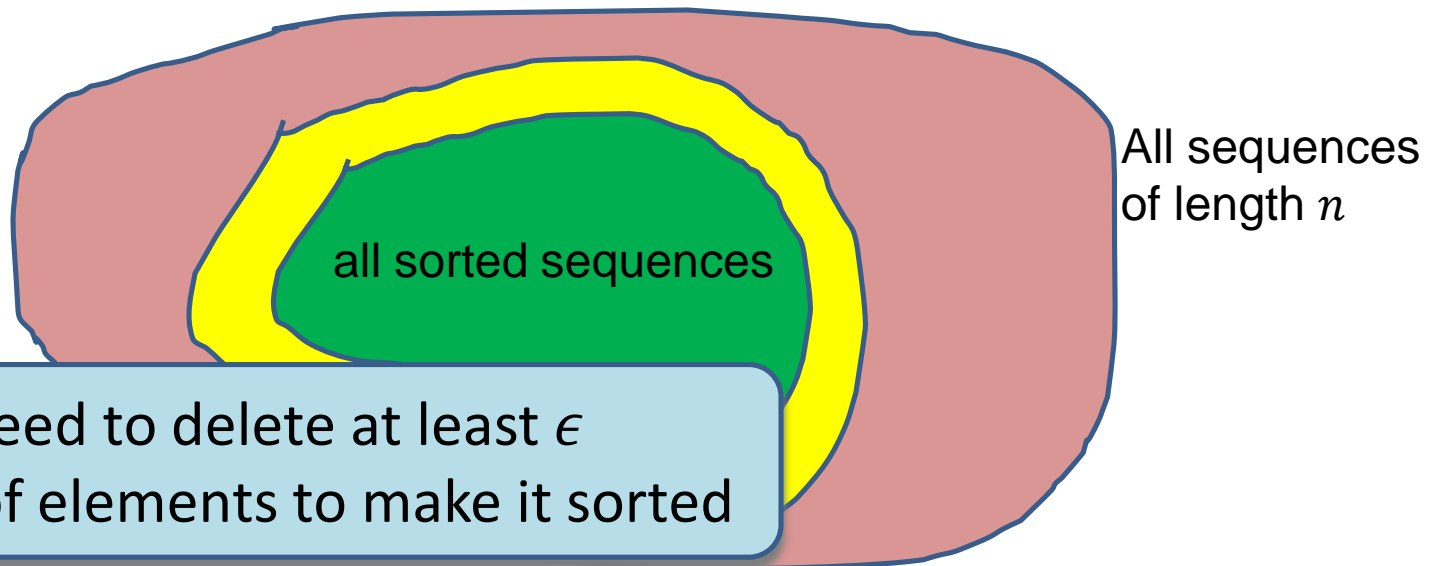
- Problem:
  - We have known distribution  $p$
  - Given samples from  $q$ , distinguish between:
    - $p = q$  vs  $\|p - q\|_1 \geq \epsilon$
    - Uniformity is an instance ( $p = U_n$ )
- Classic  $\chi^2$  test [Pearson 1900]:
  - Let  $X_i = \#$  of occurrences of  $i$
  - $\sum_i \frac{(X_i - kp_i)^2 - kp_i}{p_i} \geq \alpha$
- Test of [Valiant, Valiant 2014]:
  - $\sum_i \frac{(X_i - kp_i)^2 - X_i}{p_i^{2/3}} \geq \alpha$

# Distribution Testing++

- Other properties?
- Equality testing:
  - Given samples from unknown  $p, q$ , distinguish
  - $p = q$  vs  $\|p - q\|_1 \geq \epsilon$
  - Sample bound:  $\Theta_\epsilon(n^{2/3})$
- Independence testing:
  - Given samples from  $(p, q) \in [n] \times [n]$ , distinguish:
  - $p$  is independent of  $q$  vs  $\|(p, q) - A \times B\|_1 \geq \epsilon$  for any distributions  $A, B$  on  $[n]$
  - Sample bound:  $\widetilde{\Theta}_\epsilon(n)$
- Many more...

# Testing Monotonicity

- Problem: given a sequence  $x_1, \dots, x_n$ , distinguish:
  - sequence is sorted, vs
  - sequence is NOT sorted
- In  $o(n)$  time?
- Hard exactly: can have just one inversion somewhere
- Approximation notion:  $\epsilon$ -far



# Testing Monotonicity

$\epsilon$ -far: if need to delete at least  $\epsilon$  fraction of elements to make it sorted

- A testing algorithm:
  - Sample random positions  $i$
  - Check that  $x_i \leq x_j$  iff  $i < j$
- How many samples?
  - Bad case:  $2, 1, 4, 3, \dots, i, i-1, i+2, i+1, \dots, n, n-1$
  - At least  $\Omega(\sqrt{n})$  before we see an adjacent pair
- Fix?
  - Can sample adjacent pairs!
- Works?
  - Bad case too

# Algorithm: Monotonicity

- Assumption:
  - $x_i \neq x_j$
- One iteration:
  - Pick a random  $i$
  - Do binary search on  $x = x_i$  in the sequence
    - Start with interval  $[s,t]=[1,n]$
    - For interval  $[s,t]$ , find middle  $m = \frac{s+t}{2}$
    - If  $x < x_m$ , recurse on the left
    - If  $x > x_m$ , recurse on the right
- Fail if find inconsistency:
  - 1)  $x_i$  not found where it should be
  - 2)  $x_m \notin [x_s, x_t]$

Algorithm MONOTONICITY:

Input:  $n, x_1, \dots, x_n$

for( $r=0$ ;  $r < 3/\epsilon$ ;  $r++$ )

Let  $x = x_i$

perform binary search on  $x$

if ( $x$  not found at position  $i$

OR binary search inconsistent)

return “not sorted”;

If finished ok, return “sorted”.





# Analysis: Monotonicity

- If sorted, will pass the test
- If  $\epsilon$ -far from sorted...
  - How do we argue?
  - Via contrapositive
- **Lemma:** suppose one iteration succeeds with probability  $\geq 1 - \epsilon$ 
  - Then, sequence  $\leq \epsilon$  far from a sorted sequence
- Hence,  $3/\epsilon$  repetitions are enough to catch the case of  $> \epsilon$  far from sortedness with probability 90%:
  - Prob to report “sorted” when it’s far: is at most  $(1 - \epsilon)^{3/\epsilon} \leq e^{-3} \leq 0.1$

Algorithm MONOTONICITY:

Input:  $n, x_1, \dots, x_n$

for( $r=0$ ;  $r < 3/\epsilon$ ;  $r++$ )

Let  $x = x_i$

perform binary search on  $x$

if ( $x$  not found at position  $i$

OR binary search inconsistent)

return “not sorted”;

If finished ok, return “sorted”.



# Analysis: Monotonicity

- **Lemma:** suppose one iteration succeeds with probability  $\geq 1 - \epsilon$ 
  - Then, sequence  $\leq \epsilon$  far from a sorted sequence
- **Proof:**
  - Call  $i \in [n]$  **good** if it passes the test
  - Claim: if  $i < j$  are good, then  $x_i < x_j$ 
    - Consider the binary search tree, and their *lowest common ancestor*  $x_m$
    - It must be:
      - $x_i < x_m$  and
      - $x_m < x_j$
  - Hence: good  $i$ 's are sorted!
  - “probability  $\geq 1 - \epsilon$ ”  $\Rightarrow$  number of good elements is at least  $(1 - \epsilon)n$
  - End of proof!

Algorithm MONOTONICITY:

```
Input:  $n, x_1, \dots, x_n$ 
for( $r=0; r < 3/\epsilon; r++$ )
  Let  $x = x_i$ 
  perform binary search on  $x$ 
  if ( $x$  not found at position  $i$ 
      OR binary search inconsistent)
    return “not sorted”;
```

If finished ok, return “sorted”.

# Monotonicity: discussion

- Assumption?
  - Replace all  $x_i$  by  $(x_i, i)$
  - Then sequence must be strictly monotonic
- Test is **adaptive**:
  - Where we query depends on what we learned from the previous queries
- Do we need adaptivity?
  - No!
  - At each iteration, we query for  $x = x_i$
  - We know precisely where binary search is supposed to look at!
    - E.g., if  $i = 1$ , then it's positions:  
 $\frac{n}{2}, \frac{n}{4}, \frac{n}{8}, \dots$
  - Can generate all the positions to query at the beginning and query them all at the same time
  - Unless, binary search is inconsistent, in which case we detect this from the queries positions

Algorithm MONOTONICITY:

Input:  $n, x_1, \dots, x_n$

for( $r=0$ ;  $r < 3/\epsilon$ ;  $r++$ )

Let  $x = x_i$

perform binary search on  $x$

if ( $x$  not found at position  $i$

OR binary search inconsistent)

return “not sorted”;

If finished ok, return “sorted”.

# Monotonicity++

- $O(\log n)$  queries tight?
  - Yes
- Can consider the more general case:
  - Function  $f: \{0,1\}^d \rightarrow \{0,1\}$
  - Monotone: if  $f(x) \leq f(y)$  whenever  $x \leq y$  (coordinate-wise)
  - Can test in  $\tilde{O}_\epsilon(\sqrt{d})$  queries! [GGLRS'98, KMS'15]

# Testing graphs

- We have a graph  $G = (V, E)$ 
  - $n$  vertices
  - $m$  edges
- Dense case:
  - $m = \Theta(n^2)$
- Sparse case:
  - Degree  $d \leq O(1)$
- Property testing:
  - Eg, is graph  $G$  connected?
  - Approximation?
    - $\epsilon$ -far: if we need to delete/insert  $\geq \epsilon m$  edges

# Connectivity in sparse graph

- Approximation?
  - $\epsilon$ -far: if we need to delete/insert  $\geq \epsilon m$  edges
  - $m = dn = O(n)$
- When does it make sense?
  - $\epsilon d \ll 1$  (otherwise any sparse graph is close to being connected!)
- Assume:  $\epsilon d \ll 1$
- Algorithm:
  - For  $r = O\left(\frac{1}{\epsilon d}\right)$  times repeat:
    - Choose a random node  $s$
    - Run a BSF from  $s$
    - Until see more than  $4/\epsilon d$  node in the CC
    - If the CC is smaller, then report “disconnected”
  - Otherwise, report “connected”

# Analysis

- Claim: if  $\epsilon$ -far, then graph has at most  $\Omega(\epsilon dn)$  connected components
- Proof:
  - Suppose  $G$  has  $c$  connected component
  - Will connect, using  $O(c)$  modifications
  - Idea:
    - Just connect each connected component consecutively
    - Issue: can get higher degree than  $d$  in a CC
      - Is really an issue when all nodes in a CC have full degree
      - Just delete one edge (preserving connectivity)
- Hence, on average a CC has  $O\left(\frac{n}{\epsilon dn}\right) = O\left(\frac{1}{\epsilon d}\right)$  nodes
  - Will pick one of them with probability at least  $\epsilon d$