

# Lecture 2: Introduction to sublinear algorithms

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# Sublinear-time algorithms (examples)

#### Problem: Compute the diameter of a point set

- m points in metric space X.
- Distances given by:

$$D = \begin{bmatrix} 0 & d_{12} & \dots & d_{1m} \\ d_{21} & 0 & & \vdots \\ \vdots & & \ddots & \\ d_{m1} & \dots & & 0 \end{bmatrix}$$

- Symmetric:  $d_{ij} = d_{ji}$ - Triangle inequality:  $d_{ij} < d_{ik} + d_{kj}$
- Input size:  $n = \Theta(m^2)$

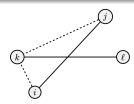


#### Algorithm 1: Diameter-Estimator

- 1 Pick k arbitrarily from  $\{1, \cdots, m\}$ ;
- **2** Pick  $\ell$  that maximizes  $\mathbf{D}_{k,\ell}$ ;
- ${\bf 3}$  Return  $k,\ell,{\bf D}_{k,\ell}$

#### Theorem

Diameter-Estimator returns a 2-approximation to the actual diameter.





# Sublinear-time algorithms (examples)

#### Problem: Number of connected components

- Input: *G* = (*V*, *E*) , |*V*| = *n*
- Goal: Estimate  $c = \ddagger$  connected components of *G*.

Let  $n_v$  :  $\sharp$  of nodes in the connected component of v.

• We need to estimate:  $c = \sum_{v \in V} \frac{1}{n_v}$ 

#### Lemma

For all  $v \in V$ , it holds that  $\frac{1}{\hat{n}_v} - \frac{1}{n_v} \le \epsilon/2$ , where  $\hat{n}_v = \min\{n_v, 2/\epsilon\}$ 



## Algorithm

Algorithm 2:  $\hat{n}_v$ -Calculator

 $\mathbf{Input} \hspace{0.1 in }: \operatorname{Graph} G, \operatorname{vertex} v, \hspace{0.1 in } \epsilon$ 

Output:  $\hat{n}_v$ .

- 1 Initialize Breadth-first search (BFS) from v;
- 2 while # of unique visited nodes by BFS is  $< \frac{2}{\epsilon}$  do
- 3 | Continue BFS ;
- 4 if BFS finishes then
- 5 Return number of visited nodes and abort

6 Return  $\frac{2}{\epsilon}$ 

#### Algorithm 3: *c*-Calculator

Input : Graph G,  $\epsilon$ , b Output:  $\tilde{c}$ . 1  $r \leftarrow b/\epsilon^3$ ; 2 Sample r vertices  $v_1, \dots, v_r$  from G uniformly with replacement; 3 Compute  $\hat{n}_{v_i}$  for all  $1 \le i \le r$  using  $\hat{n}_v$ -Calculator; 4 Return  $\tilde{c} = \frac{\pi}{r} \left( \sum_{i=1}^r 1/\hat{n}_{v_i} \right)$ 

#### Lemma

#### It holds that: $\Pr[|\hat{c} - \tilde{c}| > \epsilon n/2] \le 1/4$

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## Finishing the proof

#### Theorem

Let c be the number of connected components of G and let  $\tilde{c}$  be the output of Algorithm 3. Then,  $\Pr[|c - \tilde{c}| \le \epsilon n] \ge 3/4$ .

Proof:



Property testing definitions

### Computational problems (exact)

Search problems

- $x : R(x) = \{y : (x, y) \in R\}$
- $v: \{0, 1\}^* \rightarrow \mathbb{R}$  (value)
- Goal: Find  $y^* = \max_{y \in R} \{v(y)\}$





## Property testing definitions

### Computational problems (approximate)

Search problems

- $x : R(x) = \{y : (x, y) \in R\}$
- $v: \{0,1\}^* \rightarrow \mathbb{R}$  (value)
- Goal: Find

$$y^*: v(y^*) > C \cdot \max_{y \in R(x)} \{v(y)\}$$





## Definitions for property testing

#### Definition

Let  $\Pi_n$  be a set of functions  $f : [n] \to R_n$ ,  $n \in \mathbb{N}$ . The union  $\Pi = \bigcup_{n \in \mathbb{N}} \Pi_n \pi$  of these sets will be called a property.

- Oracle access: Query  $i \rightarrow f(i)$
- Distance: Let  $\delta(f, g) = \frac{|\{i \in [n]: f(i) \neq g(i)\}|}{n} = \Pr_{i \in U_{[n]}}[f(i) \neq g(i)]$
- Distance from property  $\Pi = \bigcup_{n \in \mathbb{N}} \Pi_n$ :
  - $\delta_{\Pi}(f) = \delta(f, \Pi) = \min_{g \in \Pi_n} \{\delta(f, g)\}$
  - $-\delta_{\Pi}(f) = \infty$  if  $\Pi_n = \emptyset$ .
- Query complexity:  $q: \mathbb{N} \times (0, 1] \rightarrow \mathbb{N}$



## Definitions for property testing

#### Definition

A tester for a property  $\pi$  is a probabilistic oracle machine that outputs a binary verdict that satisfies the following:

- 1. If  $S \in \Pi$ , then the tester accepts with probability at least 2/3.
- If S is 
  *ϵ*-far from Π, then the tester accepts with probability at most 1/3
  - One sided error: Accept any  $S \in \Pi$  with probability 1.





#### Problem: Testing convex position

#### Definition

A point set *P* is in convex position if every point in *P* belongs to the convex hull of *P*.

#### Definition

A set *P* of *n* points is  $\epsilon$ -far from convex position if no set *Q* of size (at most)  $\epsilon n$  exists such that  $P \setminus Q$  is in convex position.

**Goal:** Design a tester that can distinguish the above in sub-linear time.



### Tester

CONVEXTESTER (P,  $\epsilon$ ) let  $s = 16 \left( 4^{d+1} \sqrt{n^d/\epsilon} + 2d + 2 \right)$ Choose a set  $S \subseteq P$  of size s uniformly at random if S is in convex position then accept else reject

**Completeness case:** Clearly, if *P* is in convex position, then any  $S \subseteq P$  will also be in convex position. **Soundness case:** We need to show that **CONVEX TESTER** 

rejects every point set that is  $\epsilon$ -far from convex position with probability at least 2/3.

