

Comparing the strength of query types in Property Testing

The case of k -colorability

Ido Ben-Eliezer, Tali Kaufman, Michael Krivelevich, Dana Ron

Input graphs

Graphs of general density

$$G=(V,E), |V|=n, |E|=m$$

$d := 2m/n$ – average degree of G

$O(1) \leq d(n) \leq O(n)$ – no apriori assumptions

Measuring distance: $\text{dist}(G,H) = \text{edit.dist}(G,H)/|E(G)|$

P – property: $\text{dist}(G,P) := \min\{\text{dist}(G,H) : H \in P\}$

w.r.t. the actual size of G

- PR'02; KKR'04

Query types

- Designed to accommodate various graph densities
- degree queries: what is $d(v)$?
 - mostly convenience
- pair queries: whether $(u, v) \in E(G)$?
 - like in dense graphs/adjacency matrix
- neighbor queries: who is the i -th neighbor of v ?
 - like in bounded degree graphs/incidence lists
 - typically a random neighbor of v suffices

New query type – group query

Group query: $v \in V, S \subseteq V$

Q: whether G has an edge between v and S ?

- Motivated by group testing
- Stronger than pair query ($S=\{u\}$)
- Can be used to emulate a neighbor query in $O(\log n)$ group queries
 - \Rightarrow Essentially at least as strong as pair+neighbor queries combined
- Can recover all $d(v, S)$ edges between v and S in $O(d(v, S) \cdot \log |S|)$ queries

Main task

- Comparing strength of various query models
 - only pair queries
 - only neighbor queries
 - pair+neighbor combined
 - group queries

Test case studied: $P := k\text{-colorability}$, $k \geq 3$ constant
($k \geq 2$)

Results

	Pair Queries	Neighbor Queries	Pair&Neighbor Queries	Group Queries
Upper Bound	$\tilde{O}((\frac{n}{d})^2)$	$O(n)$	$\min\{\tilde{O}((\frac{n}{d})^2), O(n)\}$	$\tilde{O}(\frac{n}{d})$
Lower Bound	$\Omega((\frac{n}{d})^2)$	$\Omega(n^{1-\frac{1}{\lceil(k+1)/2\rceil}})$ $\Omega(n \cdot d^{-\frac{1}{\lceil k/2 \rceil - 1}})$ if $k \geq 6$	$\Omega(\frac{n}{d})$ also for 2-sided error	$\Omega(\frac{n}{d})$ also for 2-sided error

Table 1: Results for one-sided error testing of k -colorability

Interpretation, conclusions

- Group queries are at least as strong as the combined model; sometimes are strictly stronger (for $k=2$, cf. KKR'04);
- Neighbor queries are better suited for sparse graphs; increasing density does not necessarily make it easier;
- Pair queries are better suited for dense graphs;
- Combined model (pair+neighbor queries) is strictly stronger than $\text{best}(\text{pair}, \text{neighbor})$, at least for some problems.

Thank you for your (short...)
attention!