

Sublinear Time and Space Algorithms 2020B – Lecture 13

Sublinear-Time Algorithms Planar Vertex Cover (cont'd) and Communication Complexity of Equality*

Robert Krauthgamer

1 Vertex Cover in Planar Graphs via Local Partitioning (cont'd)

Recall from last week that our remaining (and main) challenge is to implement a partition oracle, i.e., an “algorithm” that can compute $P(v)$ for a queried vertex $v \in V$ in constant time. Note: P could be random, but should be “globally consistent” for the different queries v .

Algorithm Partition (used later as oracle):

Remark: It uses parameters k, ε' that will be set later (in the proof)

1. $P = \emptyset$
2. iterate over the vertices in a random order π_1, \dots, π_n
3. if π_i is still in the graph then
4. if π_i has a (k, ε') -isolated neighborhood in the current graph
5. then $S =$ this neighborhood
6. else $S = \{\pi_i\}$
7. add $\{S\}$ to P and remove S from the graph
8. output P

Definition: A (k, ε') -isolated neighborhood of $v \in V$ is a set $S \subset V$ that contains v , has size $|S| \leq k$, the subgraph induced on S is connected, and the number of edges leaving S is $e_{\text{out}}(S) \leq \varepsilon'|S|$.

Lemma 2a: Fix $\varepsilon' > 0$. Then a random vertex in G has probability at least $1 - 2\varepsilon'$ to have a $(k^*(\varepsilon'^2, d), \varepsilon')$ -isolated neighborhood.

Proof of Lemma 2a: Was seen in class, by considering the $(\varepsilon'^2, k^*(\varepsilon'^2, d))$ -partition guaranteed by Theorem 3.

Lemma 2b: For every $\varepsilon > 0$, Algorithm Partition above with parameters $\varepsilon' = \varepsilon/(12d)$ and $k = k^*(\varepsilon'^2, d)$ computes whp an (ε, k) -partition. Moreover, it can be implemented as a partition oracle (given a query vertex, it returns the part of that vertex), whose running time (and query

*These notes summarize the material covered in class, usually skipping proofs, details, examples and so forth, and possibly adding some remarks, or pointers. The exercises are for self-practice and need not be handed in. In the interest of brevity, most references and credits were omitted.

complexity into G) to answer q non-adaptive queries is whp at most $q \cdot 2^{d^{O(k)}}$.

WE SKIPPED THIS PROOF IN CLASS. But here is the rest of the proof.

Proof of Lemma 2b: By construction, the output P is a partition, where every part has size at most k .

To analyze the number of cross-edges in P , we define for each $i = 1, \dots, n$ two random variables related to π_i , as follows. Let $S_i = P(\pi_i)$, i.e. the set $S \in P$ that contains π_i (note it is removed from the graph in iteration i or earlier), and define $X_i = e_{\text{out}}'(S_i)/|S_i|$, where $e_{\text{out}}'(S_i)$ is the number of edges at the time of removing S_i . Notice that each $S \in P$ “sets” $|S|$ variables X_i to the same value, thus $\sum_i X_i = \sum_{S \in P} e_{\text{out}}'(S)$ is the number of cross-edges in P (each edge is counted once, because the graph changes with the iterations).

Now fix i . Since π_i is a random vertex, by Lemma 2a, with probability $\geq 1 - 2\varepsilon'$, it has a (k, ε') -isolated neighborhood in the input G , and also in later iterations (as that subgraph of G is planar too), in which case $X_i \leq \varepsilon'$ (both if π_i is removed in iteration i and if in an earlier iteration). With the remaining probability $\leq 2\varepsilon'$, we can bound $X_i \leq d$ which always holds. Altogether,

$$\mathbb{E}[X_i] \leq 1 \cdot \varepsilon' + 2\varepsilon' \cdot d \leq 3\varepsilon' d.$$

$$\mathbb{E}[\sum_i X_i] \leq 3\varepsilon' dn.$$

By Markov’s inequality, with probability $\geq 3/4$, the number of cross-edges in P is at most $4(3\varepsilon' dn) = \varepsilon n$.

Implementation as an oracle: We generate the permutation π on the fly by assigning each vertex v a priority $r(v) \in [0, 1]$ (and remember previously used values). Before computing $P(v)$, we first compute (recursively) $P(w)$ for all vertices w within distance at most $2k$ from v that satisfy $r(w) < r(v)$. (Note that a vertex w at distance $2k - 2$ might affect v by causing removal of a vertex mid-way between v and w .) If $v \in P(w)$ for one of them, then $P(v) = P(w)$. Otherwise, search (by brute-force) for a (k, ε') -isolated neighborhood of v , keeping in mind that vertices in any $P(w)$ as above are no longer in the graph. Searching for an optimal vertex cover inside a part is done exhaustively.

Running time: We effectively work in an auxiliary graph H , where we connect two vertices if their distance in G is at most $2k$. Thus, the maximum degree in H is at most $D = d^{2k}$. As seen earlier, this means the expected number of vertices inspected recursively is at most $D^{O(D)} = 2^{D^{O(1)}} = 2^{d^{O(k)}}$.

2 Communication Complexity

Model: Two parties, called Alice and Bob, receive inputs x, y respectively. They can exchange messages, in rounds, until one of them (or both) reports an output $f(x, y)$.

Main measure is communication complexity, i.e., total communication between the parties (in bits, worst-case).

Variants of randomization: none (deterministic), shared/public, or private.

Number of rounds: zero (simultaneous, i.e., each sends a message to a referee and not directly to each other), one (one-way communication), or more/unbounded.

Many other variants, like more players communicating in series (or broadcast etc.), with different input model (e.g., number on forehead instead of number in hand).

Equality as an Example:

Problem definition: Alice and Bob's inputs are $x, y \in \{0, 1\}^n$, and their goal is to compute $EQ(x, y) = \mathbb{1}_{\{x=y\}}$.

Public randomness: There is a (simultaneous) protocol with $O(1)$ bits.

Private randomness: There is a (one-way) protocol with $O(\log n)$ bits.

Deterministic one-way: Every protocol requires $\Omega(n)$ communication bits.