6.5240 Sublinear Time Algorithms

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Lecture 20

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1 Greedy List Coloring

Today, we discuss a sublinear time algorithm for graph coloring.

Definition 1. A proper c-coloring of a graph G = (V, E) assigns a color c_v from a palette (e.g., $\{1, \ldots, C\}$) to each $v \in V$ such that $c_u \neq c_v$ for all $(u, v) \in E$.

In this lecture, we assume that the maximum degree of a vertex in G is Δ , and $C = \Delta + 1$ (or maybe $C = 2\Delta$). Note that $(\Delta + 1)$ -coloring is easy via greedy. We call this greedy algorithm GreedyListColoring and give it in Algorithm 1. Note that GreedyListColoring takes O(|E|) time.

1 foreach $v \in V$ do 2 $L(v) \leftarrow \{1, \dots, \Delta + 1\}$ 3 foreach $v \in V$ (in an arbitrary order) do 4 if $L(v) = \emptyset$ then 5 return fail 6 else 7 $c_v \leftarrow$ any color in L(v)8 remove c_v from L(u) for all neighbors u of v

Algorithm 1: A greedy algorithm, called GreedyListColoring, for finding a $(\Delta + 1)$ -coloring in a graph G = (V, E).

The model we consider supports the following three types of queries on a graph G = (V, E):

- Degree queries: Given $u \in V$, what is deg(u)?
- Pair queries: Given $u, v \in V$, is it true that $(u, v) \in E$?
- Neighbor queries: Given $u \in V$ and $k \in \mathbb{N}$, what is the k^{th} neighbor of u?

2 Palette Sparsification

We introduce a technique called *palette sparsification* to improve the running time of graph coloring. The goal is to prove the following theorem.

Theorem 2. One can find a $(\Delta + 1)$ -coloring of an n-vertex graph in $O(n\sqrt{n})$ time.

The idea of *palette sparsification* is the following:

For each vertex $v \in V$ in an *n*-vertex graph G = (V, E), sample $k = \Theta(\log n)$ colors from $\{1, \dots, \Delta + 1\}$ to get L(v). (1)

The following lemma is the main observation, which we state without giving a proof. In Section 4, we prove a weakened version which relaxes $(\Delta + 1)$ -coloring to 2Δ -coloring.

Lemma 3. With high probability, a graph G = (V, E) can be colored by (1) such that $c_v \in L(v)$ for all $v \in V$ via GreedyListColoring.

In what follows, we denote $G_{\text{sparse}} = (V, E_{\text{sparse}})$, where $E_{\text{sparse}} = \{(u, v) \in E : L(u) \cap L(v) \neq \emptyset\}$. Figure 1 gives an example of G_{sparse} .



Figure 1: An example of palette sparsification, where edges absent from G_{sparse} are crossed off, and each vertex is colored by the colors in its palette.

It turns out that G_{sparse} does not contain many edges.

Lemma 4. With high probability, $|E_{sparse}| = O(n \log^2 n)$.

Proof. Fix $u \in V$. Without loss of generality, suppose that $L(u) = \{1, \ldots, k\}$. For all $v \in N(u)$ and $i \in \{1, \ldots, k\}$, set

$$X_{v,i} = \begin{cases} 1, & \text{if } i \in L(v), \\ 0, & \text{otherwise.} \end{cases}$$

Let

$$X = \sum_{i=1}^{k} \sum_{v \in N(u)} X_{v,i}.$$

Since $\sum_{v \in N(u)} X_{v,i}$ is the number of edges due to color *i*, then X is an upper bound on deg(*u*) in G_{sparse} . Note that $\mathbb{E}[X_{v,i}] = k/(\Delta + 1)$ for all $v \in N(u)$ and $i \in \{1, \ldots, k\}$. Hence,

$$\mathbb{E}[X] = \mathbb{E}\left[\sum_{i=1}^{k} \sum_{v \in N(u)} X_{v,i}\right] = \sum_{i=1}^{k} \sum_{v \in N(u)} \mathbb{E}\left[X_{v,i}\right] \le k \cdot \Delta \cdot \frac{k}{\Delta + 1} < k^2.$$

Since $k = \Theta(\log n)$, then $\mathbb{E}[\deg(u)] \leq O(\log^2 n)$. (One can show "with high probability" with additional work.)

3 A Sublinear Time Algorithm for Graph Coloring

Assuming Lemma 3, we give a sublinear time algorithm for finding a $(\Delta + 1)$ -coloring of a graph:

- 1. Construct the palette of each vertex using (1).
- 2. Construct G_{sparse} : For each color $c \in \{1, \ldots, \Delta + 1\}$, find $X_c = \{v \in V : c \in L(v)\}$ (do this while doing step 1). Query all pairs of vertices in each X_c to find E_{sparse} : suppose that $X_c = \{v_{i_1}, \ldots, v_{i_\ell}\}$; for distinct $j, k \in [\ell]$, query if $(v_{i_j}, v_{i_k}) \in E$, and if so, add it to E_{sparse} .
- 3. Perform GreedyListColoring on G_{sparse} .

Step 1 takes $O(n \log n)$ time. Step 3 takes $O(|E_{\text{sparse}}|) = O(n \log^2 n)$ time. For each color c,

$$\mathbb{E}\left[|X_c|^2\right] = \sum_{\substack{u,v \in V \\ u \neq v}} \mathbb{E}\left[\mathbbm{1}_{u,v \text{ both choose color } c}\right] = \binom{n}{2} \left(\frac{k}{\Delta+1}\right)^2 = O\left(\frac{n^2 \log^2 n}{\Delta^2}\right).$$

Hence, the running time to query all pairs in each X_c is at most

$$(\Delta + 1) \cdot O\left(\frac{n^2 \log^2 n}{\Delta^2}\right) = \tilde{O}\left(\frac{n^2}{\Delta}\right).$$

It follows that the total running time is $\tilde{O}(n^2/\Delta)$. This proves Theorem 2.

Proof of Theorem 2. If $\Delta \leq \sqrt{n}$, then we use GreedyListColoring with $O(|E|) \leq O(n\Delta) \leq O(n\sqrt{n})$ time. If $\Delta > \sqrt{n}$, then we use palette sparsification with $\tilde{O}(n^2/\Delta) \leq \tilde{O}(n^2/\sqrt{n}) = \tilde{O}(n^{3/2})$ time. \Box

4 Relaxing Lemma 3 to 2Δ -Coloring

In this section, we prove a weakened version of Lemma 3 which relaxes $(\Delta + 1)$ -coloring to 2Δ coloring. Other parts in the proof of Theorem 2 remain the same, hence giving an $\tilde{O}(n^{3/2})$ time algorithm for finding a 2Δ -coloring of an *n*-vertex graph.

Lemma 5. With high probability, a graph G = (V, E) can be colored by (1) with $\Delta + 1$ replaced by 2Δ such that $c_v \in L(v)$ for all $v \in V$ via GreedyListColoring.

Proof. When attempting to color a vertex v, we say that a color $c \in \{1, \ldots, 2\Delta\}$ is good if $c \in L(v)$ initially and c is not used to color any *previous* neighbor of v. If L(v) contains any good color c, then we can color v successfully. Since L(v) is chosen independently of its neighbors, we can think of choosing L(v) "now." Since v has at most Δ neighbors, then

$$\Pr[L(v) \text{ contains no good color}] \le \frac{\binom{\Delta}{k}}{\binom{2\Delta}{k}} = \frac{\frac{\Delta(\Delta-1)\cdots(\Delta-k+1)}{k!}}{\frac{(2\Delta)(2\Delta-1)\cdots(2\Delta-k+1)}{k!}} < \frac{1}{2^k} = \left(\frac{1}{2}\right)^{\Theta(\log n)} = \frac{1}{n^{\alpha}},$$

for some constant α . By the union bound,

Pr[there exists a vertex v such that L(v) has no good color] $\leq \frac{1}{n^{\alpha'}}$,

for some constant α' (e.g., $\alpha' = 3$). Hence, with high probability, the algorithm never fails. It follows that G has a legal list coloring with high probability.

We note that the proof of Lemma 5 indeed generalizes to $(1 + \delta)\Delta$ -coloring for any constant $\delta > 0$ which does not depend on Δ .