

3 Deterministic vs Randomized complexities

Theorem 0.1 (Linial [Lin92]). *Let \mathbf{G} be the class of all graphs of degree bounded by Δ and Π be the proper vertex $(\Delta + 1)$ -coloring problem. Then we have*

$$\Pi \in \text{LOCAL}(O(\log^* n)).$$

Proof Sketch. The proof is based on an iterated application of the following combinatorial result on *cover free set systems*.

Claim 0.2 (Section 4.8.2 in [?]). *For every $n > \Delta > 2$ there is $m \leq 4(\Delta + 1)^2 \log^2(n)$ and a family of subsets $(S_i)_{i \in [n]}$ of $[m]$ such that*

$$S_{i_0} \not\subseteq \bigcup_{\ell \in [\Delta]} S_{i_\ell},$$

whenever $i_0 \neq i_\ell$ for every $\ell \in [\Delta]$.

Start with a graph G on n vertices with unique identifiers from $\text{poly}(n)$ and then use Claim 0.2 iteratively to decrease the colors to $\text{poly}(\Delta)$ in $\log^* n$ rounds as follows. Suppose that we have coloring c_x with x -colors and x is larger than e.g. $(11(\Delta + 1))^3$, see [? , Section 4.8.3]. Pick m and $(S_i)_{i \in [n]}$ as in Claim 0.2. In one communication round let $c'(v)$ be the lexicographically minimal color in

$$S_{c_x(v)} \setminus \bigcup_{(v,w) \in E(G)} S_{c_x(w)}.$$

Then c' is a well defined m -coloring. It can be shown that as long as $\log^{(i)} x \geq \log 36 + 2 \log(\Delta + 1)$, then we decrease the number of colors from

$$(6(\Delta + 1) \log^{(i-1)} x)^2 \text{ to } (6(\Delta + 1) \log^{(i)} x)^2$$

in the i -th step, where $i > 2$.

Last step, as we treat Δ as a constant, is to use the greedy algorithm. In general, this takes as many rounds as we have colors, in our case $\text{poly}(\Delta)$. \square

Another general deterministic speed up is from $o(\log \log^* n)$ to $O(1)$, [NS95, CP17].

Randomness. Every deterministic local algorithm can be simulated as a randomized algorithm of the same complexity.

Claim 0.3. *Let $\Pi \in \text{LOCAL}(O(f(n)))$. Then $\Pi \in \text{RLOCAL}(O(f(n)))$.*

Proof. Let $m \geq n$ and split $[0, 1]$ into m -intervals $(I_j)_j$ of the same size. Let $v \neq w \in V(G)$. Then the probability that their labels do not differ is $1/m$. By the union bound, we have that the probability of having a pair with a label in the same interval is at most $\frac{n^2}{m}$. Consequently, taking $m = n^3$ yields the result. \square

Results in the opposite direction are called *derandomization*. It is known that this can be done in the regime $o(\log \log n)$, [CKP16]. In particular cases, e.g., grids this can be improved to $o(\log n)$.

Proposition 0.4 (Corollary 1 in [?]). *Let \mathbf{G} be closed under disjoint unions of finite graphs, or suppose that each $G \in \mathbf{G}_n$ can be found as an induced subgraph of some graph in \mathbf{G}_m for every $m > n$. Then*

$$\text{RLOCAL}(O(f(n))) = \text{LOCAL}(O(f(n)))$$

for every $f \in O(\log^* n)$.

Proof Sketch. Let \mathcal{A} be a randomized local algorithm of complexity $(r_n)_n \in O(\log^* n)$ that solves an LCL Π of locality $t \in \mathbb{N}$. Fix $n \in \mathbb{N}$. The idea is to find a (deterministic) function $\varphi : [n] \rightarrow [0, 1]$ and $m(n) > n$ such that running \mathcal{A} for $r_{m(n)}$ many rounds on any $G \in \mathbf{G}_n$ labeled with images of unique identifiers under $\varphi(-)$ produces Π -coloring. This will certainly produce a deterministic local algorithm of complexity $(r_{m(n)})_n$.

We show that $m(n) = 2^{n^2}$ works as required. Note that in that case, if $f \in O(\log^* n)$, then $f(m(n)) \in O(f(n))$. Let $\mathcal{G}_n \in \mathbf{G}_{m(n)}$ be a graph that contains as induced subgraph the graph \mathcal{H}_n that is created by taking disjoint copies of all graphs from \mathbf{G}_n labeled with unique identifiers. Note that $|\mathcal{H}_n| \ll m(n)$. Pick φ uniformly at random. By the definition the probability of the event that $\mathcal{A}_{m(n)}$ fails at a given vertex $v \in V(G)$ for some $G \in \mathbf{G}_n$ (labeled with images of unique identifiers under φ) is at most $\frac{1}{m(n)}$ by the definition. As $|\mathcal{H}_n| \ll m(n)$ we get by the union bound that there exists φ that does not fail at any vertex. \square

In general, randomness does help in the regime $o(\log n)$. **Picture.** For example, deterministic complexity of the *sinkless orientation* problem on trees is $\Omega(\log n)$, however, randomized complexity is $O(\text{poly}(\log \log n))$. This is the most significant complexity class that is tightly connected with the distributed complexity of solutions to the *Lovász Local Lemma (LLL)*. On a high level, given a randomized algorithm \mathcal{A} that solves Π in $o(\log n)$ rounds. We consider the *bad events* around vertices, when \mathcal{A} fails. Then fixing k large enough we try to avoid bad events of \mathcal{A} when applied to graphs of size k . Note that uniformly the probability of bad event p is less than $\frac{1}{k}$ and the degree d of bad event in the auxiliary graph of bad events is at most $\Delta^{o(\log k)} \ll k$. That is the LLL condition $p \cdot d \ll 1$ is satisfied. More on this later.

References

- [CKP16] Yi-Jun Chang, Tswi Kopelowitz, and Seth Pettie. An exponential separation between randomized and deterministic complexity in the LOCAL model. In *Proc. 57th IEEE Symp. on Foundations of Computer Science (FOCS)*, 2016.
- [CP17] Yi-Jun Chang and Seth Pettie. A time hierarchy theorem for the LOCAL model. In *Proc. 58th IEEE Symp. on Foundations of Computer Science (FOCS)*, pages 156–167, 2017.
- [Lin92] Nati Linial. Locality in distributed graph algorithms. *SIAM Journal on Computing*, 21(1):193–201, 1992.
- [NS95] Moni Naor and Larry Stockmeyer. What can be computed locally? *SIAM Journal on Computing*, 24(6):1259–1277, 1995.