# Notes for Luby Algorithm: Parallel Maximal Independent Sets

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January 28, 2016

## 1 The Algorithm

Problem : Given a graph find a maximal independent set.

- 1. I =  $\emptyset$ , G'= G.
- 2. While (G' is not the empty graph) do IN PARALLEL
  - (a) Choose a random set of vertices  $S \in G'$  by selecting each vertex v independently with probability 1/2d(v).
  - (b) For every edge  $(u, v) \in E(G')$  if both endpoints are in S then remove the vertex of lower degree from S (Break ties arbitrarily). Denote this new set S'.
  - (c)  $I = I \cup S'$ . G' = G'  $(S' \cup N(S'))$ , i.e., G' is the induced subgraph on V'  $(S' \cup N(S'))$  where V' is the previous vertex set.

3. output the independent set I

**Correctness :** We see that at each stage the set S' that is added is an independent set. Moreover since we remove, at each stage,  $S' \cup N(S')$  the set I remains an independent set. Also note that all the vertices removed from G' at a particular stage are either vertices in I or neighbors of some vertex in I. So the algorithm always outputs a maximal independent set.

- A single round can be done in constant time using  $O(|V|^2)$  processors
- The expected value of the number of rounds is in  $O(\log n)$ . With n = |E|

## 2 Expected Running Time (Number of rounds)

In this section we prove that the algorithm take O(log n) time. Let  $G_j = (V_j, E_j)$  denote the graph after stage j

Main Lemma: For some k < 1,

$$E(|E_j| | E_{j-1}) < k|E_{j-1}|.$$

Hence, in expectation, only  $O(\log n)$  rounds will be required, where  $n = |E_0|$ . We say vertex v is *bad* if more than 2/3 of the neighbors of v are of higher degree than v. We say an edge is *bad* if both of its endpoints are bad, otherwise the edge is *good*.

The key claims are that at least half the edges are *good*, and each *good* edge is deleted with a constant probability. The main lemma then follows immediately.

Lemma 1 : At least half the edges are good.

**Proof**: Denote the set of bad edges by  $E_B$ . We will define  $f: E_B \to \begin{pmatrix} E \\ 2 \end{pmatrix}$  so that for all  $e_1 \neq e_1 \in E_B, f(e_1) \cap f(e_2) = \emptyset$ . This proves  $|E_B| \leq |E|/2$ , and we are done.

The function f is defined as follows. For each  $(u, v) \in E$ , direct it to the higher degree vertex. Break ties as in the algorithm. Now, suppose  $(u, v) \in E_B$ , and is directed towards v. Since v is *bad*, it has at least twice as many edges out as in. Hence we can pair two edges out of v with every edge into v.

**Lemma 2:** If v is good then  $Pr(N(v) \cap S \neq \emptyset) \geq 2\alpha$ , where  $\alpha = 1/2 \times (1 - e^{-1/6})$ .

**Proof:** Define  $L(v) := \{w \in N(v) | d(w) \le d(v) \}$ .

By definition,  $|L(v)| \ge d(v)/3$  if v is a good vertex.

$$\begin{split} \Pr\left( \left. N(v) \cap S \neq \emptyset \right) &= 1 - \Pr\left( \left. N(v) \cap S = \emptyset \right) \\ &= 1 - \prod_{w \in N(v)} \Pr\left( w \notin S \right) \quad \text{using full independence} \\ &\geq 1 - \prod_{w \in L(v)} \Pr\left( w \notin S \right) \\ &= 1 - \prod_{w \in L(v)} \left( 1 - \frac{1}{2d(w)} \right) \\ &\geq 1 - \prod_{w \in L(v)} \left( 1 - \frac{1}{2d(v)} \right) \\ &\geq 1 - \exp(-|L(v)|/2d(v)) \\ &\geq 1 - \exp(-|L(v)|/2d(v)) \\ &\geq 1 - \exp(-1/6), \end{split}$$

Note, the above lemma is using full independence in its proof. And Pr: Probability.

 $\label{eq:Lemma 3: Pr} \begin{array}{l} \mbox{Ir} (\ w \in S' \mid w \in S \ ) \leq 1/2. \end{array}$   $\label{eq:Proof: Let H(w) = N(w) \setminus L(w) = \{z \in N(w) : d(z) > d(w)\}. \end{array}$ 

$$\begin{split} \Pr\left(w \notin S' \mid w \in S\right) &= \Pr\left(H(w) \cap S \neq \emptyset \mid w \in S\right) \\ &\leq \sum_{z \in H(w)} \Pr\left(z \in S \mid w \in S\right) \\ &\leq \sum_{z \in H(w)} \frac{\Pr\left(z \in S, w \in S\right)}{\Pr\left(w \in S\right)} \\ &= \sum_{z \in H(w)} \frac{\Pr\left(z \in S\right) \Pr\left(w \in S\right)}{\Pr\left(w \in S\right)} \text{ using pairwise independence} \\ &= \sum_{z \in H(w)} \Pr\left(z \in S\right) \\ &= \sum_{z \in H(w)} \frac{1}{2d(z)} \\ &\leq \sum_{z \in H(w)} \frac{1}{2d(v)} \\ &\leq \frac{1}{2}. \end{split}$$

**Lemma 4:** If v is good then  $\Pr( v \in N(S') ) \geq \alpha$ 

**Proof:** Let  $V_G$  denote the *good* vertices. We have

$$\begin{aligned} \Pr\left( v \in N(S') \mid v \in V_G \right) &= \Pr\left( N(v) \cap S' \neq \emptyset \mid v \in V_G \right) \\ &= \Pr\left( N(v) \cap S' \neq \emptyset \mid N(v) \cap S \neq \emptyset, v \in V_G \right) \Pr\left( N(v) \cap S \neq \emptyset \mid v \in V_G \right) \\ &\geq \Pr\left( w \in S' \mid w \in N(v) \cap S, v \in V_G \right) \Pr\left( N(v) \cap S \neq \emptyset \mid v \in V_G \right) \\ &\geq (1/2)(2\alpha) \\ &= \alpha \end{aligned}$$

**Corollary 4:** If v is *good* then the probability that v gets deleted is at least  $\alpha$ .

**Corollary 5:** If an edge e is *good* then the probability that it gets deleted is at least  $\alpha$ .

**Proof:** Pr (  $\mathbf{e} = (\mathbf{u}, \mathbf{v}) \in E_{j-1} \setminus E_j \ge \Pr$  (  $\mathbf{v}$  gets deleted ).

We now return the main lemma :

### Main Lemma :

$$E(|E_j| | E_{j-1}) < k|E_{j-1}|(1 - \alpha/2)$$

**Proof:** 

$$\begin{array}{lll} {\rm E} \left( \; |E_j| \; \mid \; E_{j-1} \; \right) & = & \sum_{e \in E_{j-1}} 1 - \Pr \left( \; e \; {\rm gets \; deleted} \; \right) \\ & \leq & |E_{j-1}| - \alpha |GOOD \; {\rm edges}| \\ & \leq & |E_{j-1}| (1 - \alpha/2). \end{array}$$

The constant  $\alpha$  is approximately 0.076.

Thus,

$$E(|E_j|) \le |E_0|(1-\alpha/2)^j \le n \exp(-j\alpha/2) < 1$$

for j ; 2/ $\alpha$  log n. Therefore, the expected number of rounds required is  $\leq 4n = O(\log n)$ .

#### Theorem:

The expected number of rounds in Luby algorithm for finding maximal independent set is in O(log n).

### References

- Motwani, Rajeev, and Prabhakar Raghavan. Randomized algorithms. Chapman and Hall/CRC, 2010.
- [2] Maximal independent set https://en.wikipedia.org/wiki/Maximal\_independent\_set.
- [3] Blelloch, Guy E., Jeremy T. Fineman, and Julian Shun. "Greedy sequential maximal independent set and matching are parallel on average." Proceedings of the twenty-fourth annual ACM symposium on Parallelism in algorithms and architectures. ACM, 2012.
- [4] Luby, M. (1986). "A Simple Parallel Algorithm for the Maximal Independent Set Problem". SIAM Journal on Computing 15 (4): 1036.
- [5] Luby's Algorithm, in: Lecture Notes for Randomized Algorithms, Last Updated by Eric Vigoda on February 2, 2006