Math 261C: Randomized Algorithms

Lecture topic: WalkSat, part II & Lovász Local Lemma, part I

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1. WALKSAT, PART II

We now analyze the WalkSat algorithm described in the previous lecture for values of k at least 3.

Recall that the WalkSat algorithm can be compared to a Markov process on the states $\{0, 1, 2, \ldots, n\}$, where n is the number of variables in the given instance of k-SAT. These states represent the "distance" from the algorithm's current generated assignment to the nearest satisfying assignment. Let q_j be the probability of reaching state 0 in at most m = 3n steps starting from the state j.

We will allow *i* "rightward" (away from 0) moves before ending up at state 0. This requires j+2i moves, which is less than 3n since $i, j \in \{0, 1, 2, ..., n\}$. The probability of a rightward move is $\frac{k-1}{k}$ and the probability of a leftward move is $\frac{1}{k}$.

Therefore we have

$$q_j \ge P$$
 [ending up at 0 after at most $j + 2i$ moves]
 $\ge {\binom{j+2i}{i}} \left(\frac{1}{k}\right)^{j+i} \left(\frac{k-1}{k}\right)^i.$

Let $i = \left\lceil \frac{j}{k-2} \right\rceil$ so that $i \approx \alpha j$ for $\alpha = \frac{1}{k-2}$. Then $j + 2i \approx \beta j$ for $\beta = 1 + 2\alpha = \frac{k}{k-2}$.

Then by Stirling's Formula (which states $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$) we have

$$\begin{pmatrix} j+2i\\i \end{pmatrix} \approx \begin{pmatrix} \beta j\\\alpha j \end{pmatrix}$$

$$= \frac{(\beta j)!}{(\alpha j)!((\beta - \alpha)j)!}$$

$$\approx \frac{1}{\sqrt{2\pi}} \sqrt{\frac{\beta}{\alpha(\beta - \alpha)}} \frac{1}{\sqrt{j}} \left(\frac{\beta^{\beta}}{\alpha^{\alpha}(\beta - \alpha)^{\beta - \alpha}}\right)^{j}.$$

Now $\beta - \alpha = \frac{k-1}{k-2}$. So, ignoring the constants out in front, we can say that the order of the binomial coefficient in question is

$$\binom{j+2i}{i} \approx \frac{1}{j^{\Theta(1)}} \left(\frac{k^{k/k-2}}{(k-1)^{(k-1)/(k-2)}} \right)^j.$$

Using this in our lower bound for q_j we find

$$q_j > \frac{1}{j^{\Theta(1)}} \left(\frac{1}{k-1}\right)^j.$$

Now recall that the algorithm for WalkSat has the form

Algorithm: WalkSat(Γ , k, n)

Loop: Choose at random a truth assignment for the variables.

Loop (at most 3n times): Check if formula is satisfied.

If not, flip a random literal in a random unsatisfied clause.

The probability p that the inner loop finds a satisfying assignment (assuming one exists) can be estimated by

$$\begin{split} p &> \sum_{j=0}^{n} P\left[\text{we disagree from a satisfying assignment in } j \text{ values}\right] q_j \\ &\geq \sum_{j=0}^{n} \left(\frac{1}{2^n} \binom{n}{j}\right) \frac{1}{n^{\Theta(1)}} \left(\frac{1}{k-1}\right)^j \\ &= \frac{1}{n^{\Theta(1)}} \frac{1}{2^n} \sum_{j=0}^{n} \binom{n}{j} \left(\frac{1}{k-1}\right)^j \\ &= \frac{1}{n^{\Theta(1)}} \frac{1}{2^n} \left(1 + \frac{1}{k-1}\right)^n \qquad \text{by the binom} \\ &= \frac{1}{n^{\Theta(1)}} \frac{1}{2^n} \left(\frac{k}{k-1}\right)^n \\ &= \frac{1}{n^{\Theta(1)}} \left(\frac{1}{2} \frac{k}{k-1}\right)^n \\ &> \frac{1}{2^n}, \end{split}$$

nial theorem

and so the expected number of iterations of the outer loop before it succeeds is approximately

$$n^{\Theta(1)} \left(2\frac{k-1}{k}\right)^n < 2^n.$$

This is an improvement over the previous algorithm, but still runs in exponential time.

Below is a table for the expected runtime of WalkSat for various values of k.



Compare this to the conjectured runtimes for deterministic algorithms for SAT.

Exponential Time Hypothesis: Let $k \ge 3$. Any deterministic algorithm for SAT takes time $n^{\Theta(1)}(s_k)^n$ in k-SAT instances for some constant $s_k > 1$.

Strong Exponential Time Hypothesis: For any deterministic algorithm for SAT, $\lim_{k\to\infty} s_k = 2$.

For a derandomization of the WalkSat algorithm see [1].

2. Lovász Local Lemma, Part I

Definition. Let E_1, E_2, \ldots, E_n be events in some probability space. We say G is a dependency graph for E_1, \ldots, E_n if G has vertex set $\{1, 2, \ldots, n\}$ and for all $i E_i$ is independent of the set $\{E_i : \text{there is no edge from } i \text{ to } j\}$.

We will use the notation

 $\Gamma_i = \{j : \text{ there is an edge from } i \text{ to } j\}.$

Lovász Local Lemma. Let E_1, \ldots, E_n and G be as in the definition. Also let $x_1, x_2, \ldots, x_n \in [0, 1)$ be such that

$$P[E_i] \le x_i \prod_{j \in \Gamma_i} (1 - x_j).$$

Then

$$P[\overline{E_1} \wedge \ldots \wedge \overline{E_n}] \ge \prod_{i=1}^n (1 - x_i).$$

Corollary. Suppose G has degree at most d (i.e., for all i we have $|\Gamma_i| \leq d$). Suppose also that $P[E_i] \leq p$ for all i. If either

$$ep(d+1) \leq 1 \text{ or } 4pd \leq 1$$

then

$$P(\overline{E_1} \wedge \ldots \wedge \overline{E_n}) > 0.$$

Proof of the Corollary. Let $x_i = \frac{1}{d+1}$ for all *i*. To apply the Lovász Local Lemma we need to show that

$$p \le \frac{1}{d+1} \left(1 - \frac{1}{d+1} \right)^d.$$

This will show that $P(E_i) \leq x_i \prod_{j \in \Gamma_i} (1 - x_j)$ and it will follow that $P(\overline{E_1} \land \ldots \land \overline{E_n}) > 0$.

In the first case, if $ep(d+1) \leq 1$, it suffices to show that $\frac{1}{e} \leq \left(1 - \frac{1}{d+1}\right)^d$. We can show this easily by taking the logarithm of both sides.

Otherwise, if $4pd \leq 1$, it suffices to show that

$$\frac{1}{4d} \le \frac{1}{d+1} \left(1 - \frac{1}{d+1} \right)^d = \frac{1}{d+1} \left(\frac{d}{d+1} \right)^d.$$

This is true if and only if

$$\frac{1}{4} \le \left(\frac{d}{d+1}\right)^{d+1} = \left(1 - \frac{1}{d+1}\right)^{d+1}$$

Taking logarithms, we need to verify that

$$-\log(4) \le (d+1)\log\left(1 - \frac{1}{d+1}\right),$$

which is true for all $d \ge 1$. Thus the corollary is proved.

For an improved, and optimal, version of the Corollary, see [2].

References

[1] Moser, R.A. and D. Scheder. A Full Derandomization of Schoening's k-SAT Algorithm. SToC 2011.

[2] Shearer. On a Problem of Spencer. Combinatorica 5 (1985) 241-245.