Lecture 9

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## 1 Brief Introduction to Second Moment Method

In previous lecture, we presented a lower bound for threshold of satisfiability problem by *unit clause propagation algorithm*. In this lecture, we use the *second moment method* to give another, and better lower bound for K-SAT problem. We recall the basic setting below.

Let  $\mathbf{P}_N(k, \alpha)$  be the probability that a random formula from  $SAT_N(k, M = \alpha N)$  ensemble is SAT. we try to find a lower bound of  $\alpha$ , such that  $P_N(k, \alpha) \to 0$  as  $N \to \infty$ .

At first, let's go over the main ideas of second moment method. Given a nonnegative function U, which is defined in the space of all K-SAT formulas, satisfying  $U(\psi) = 0$  if  $\psi$  is not SAT. Then,

$$\mathbf{P}[\psi \text{ is SAT}] \ge \mathbf{P}[U(\psi) > 0] \ge \frac{\mathbf{E}^2[U(\psi)]}{\mathbf{E}[U^2(\psi)]},$$

by Cauchy-Shwarz inequality. So, we try to pick up some function U whose moments can be evaluated efficiently, and give us a meaningful bound as well.

**Remark 1** It is an art to pick up U. For example, if we take  $U(\psi) = \#$  of satisfying assignments, it will give us nothing but a zero lower bound.

Here, we take

$$U(\psi) = \sum_{x \in \{0,1\}^N} \prod_{a=1}^M \omega(x,a),$$

in which

$$\omega(x,a) = \begin{cases} 0, & \text{if } C_a(x) = 0; \\ \lambda^{r(x,a)}, & \text{if } C_a(x) = 1, r(x,a) = \# \text{ of variables in } x \text{ satisfying } C_a. \end{cases}$$

Now, we try to computer the first and second moment of  $U(\psi)$ .

## 2 Computation of First Moment

By the definition of U,

$$\mathbf{E}[U(\psi)] = \mathbf{E}\left[\sum_{x \in \{0,1\}^N} \prod_{a=1}^M \omega(x,a)\right]$$
$$= \sum_{x \in \{0,1\}^N} \mathbf{E}\left[\prod_{a=1}^M \omega(x,a)\right]$$
$$= 2^N \mathbf{E}[\omega(x,a)]^M$$

Last equality holds since the clauses in a formula are chosen independently (with repetition). Since

$$\mathbf{E}[\omega(x,a)] = 2^{-k} \sum_{r=1}^{k} \binom{n}{k} \lambda^{r}$$
$$= 2^{-k} [(1+\lambda)^{k} - 1]$$

We conclude:

Claim 2 
$$\frac{\log \mathbf{E}[U(\psi)]}{N} = \log 2 - \alpha k \log 2 + \alpha \log[(1+\lambda)^k - 1] \equiv h_1(\lambda, \alpha).$$

## 3 Computation of Second Moment

The second moment calculations are typically more involved:

$$\begin{split} \mathbf{E}[U^2(\psi)] &= \sum_{x,y \in \{0,1\}^N} \mathbf{E}[\prod_{a=1} M\omega(x,a) \prod_{a=1} M\omega(y,a)] \\ &= \sum_{x,y \in \{0,1\}^N} \{ \mathbf{E}[\omega(x,a)\omega(y,a)] \}^M \end{split}$$

 $(let L = d_H(x, y) Hamming distance)$ 

$$= \sum_{x \in \{0,1\}^{N}} \sum_{L=0}^{N} \sum_{y:d_{H}(y,x)=L}^{N} \mathbf{E} \{ [\omega(x,a)\omega(y,a)] \}^{M}$$
$$= 2^{N} \sum_{L=0}^{N} {N \choose L} (g(N,L))^{M},$$
(1)

where

$$g(N,L) = 2^{-k} \sum_{\substack{u,v \neq 0, \\ u,v \in \{0,1\}^k}} \lambda^{w(u)} \lambda^{w(v)} (\frac{L}{N})^{d(u,v)} (1 - \frac{L}{N})^{k-d(u,v)}$$

Here, w(u) denotes # of 1's in u. (1) is true because

Claim 3 If  $d_H(x, y) = L$ , then  $\mathbf{E}[\omega(x, a)\omega(y, a)] = g(N, L)$ .

**Proof:** Denote u as  $(u_1, u_2, \ldots, u_k)$ , and let  $u_i = 1$  if and only if the *i*th position of a is satisfied by x, i.e., the *i*th variable in a and the counterpart in x have the same sigh. And similarly define v. Now we fix u, v and then count the number of corresponding clauses. We give a brief case by case analysis. If  $u_i = v_i$ , then the *i*th variable of a can be and only be picked up from those for which x and y have the same value; if  $u_i \neq v_i$ , then the *i*th variable of a can be and only be picked up from those for which x and y have the same value; if  $u_i \neq v_i$ , then the *i*th variable of a can be and only be picked up from those for which x and y have the same value. The values of a can be and only be picked up from those for which x and y have different values. And in both cases, the sigh of the variable has one and only one choice. Summing over all possible choices of u and v, we get the equality in the claim.  $\Box$ 

Now, we turn to the computation of g(N, L). For simplicity of notation, we let  $Z = \frac{L}{N}$ . So,

$$2^{k}g(N,L) = \sum_{u,v \in \{0,1\}^{k}} \lambda^{w(u)+w(v)} Z^{d_{H}(u,v)} (1-Z)^{k-d_{H}(u,v)} -2 \sum_{u \in \{0,1\}^{k}} \lambda^{w(u)} Z^{w(u)} (1-Z)^{k-w(u)} + (1-Z)^{k} \equiv I_{1} - 2I_{2} + (1-Z)^{k}.$$
(2)

By fixing the Hamming distance of u and v and summing over the number of 1's in common positions, we get

$$I_{1} = \sum_{d=0}^{k} \binom{k}{d} 2^{d} Z^{d} (1-Z)^{k-d} \sum_{r=0}^{k-d} \lambda^{r+d} \lambda^{r} \binom{k-d}{r}$$
$$= \sum_{d=0}^{k} \binom{k}{d} 2^{d} Z^{d} \lambda^{d} (1-Z)^{k-d} (1+\lambda^{2})^{k-d}$$
$$= (2\lambda Z + (1-Z)(1+\lambda^{2}))^{k}.$$
(3)

And fixing number of 1's in u and then summing over, we get

$$I_{2} = \sum_{r=0}^{k} {k \choose r} \lambda^{r} Z^{r} (1-Z)^{k-r} = (\lambda Z + (1-Z))^{k}.$$
(4)

Combining (2), (3) and (4), we get

$$2^{k}g(N,L) = (2\lambda Z + (1-Z)(1+\lambda^{2}))^{k} - 2(\lambda Z + (1-Z))^{k} + (1-Z)^{k}.$$

Based on discussions above, we are ready to make a claim as follow:

Claim 4  $\log \mathbf{E}[U^2(\psi)]/N \approx \log 2 + \max_{0 \leq Z \leq 1} \{-Z \log Z - (1-Z) \log(1-Z) + \alpha \log f(Z, \lambda)\} - \alpha k \log 2 \equiv h_2(\lambda, \alpha, Z), in which <math>f(Z, \lambda) = 2^k g(N, L).$ **Proof:** Using Stirling formula for the factorials and note that the order of logarithm of sum depends mainly on the maximum term in the sum.  $\Box$ 

## 4 Comparisons of Two Moments and Conclusion

We summarize the remaining steps of the argumet without detail.

• 
$$h_2(\lambda, \alpha, \frac{1}{2}) = 2h_1(\lambda, \alpha).$$

- Unless  $Z = \frac{1}{2}$  maximizes  $h_2(\lambda, \alpha, \frac{1}{2}), \frac{\mathbf{E}^2[U]}{\mathbf{E}[U^2]}$  is exponentially small when N goes to  $\infty$ .
- $Z = \frac{1}{2}$  being a maximizer implies that  $(1 + \lambda)^{k-1} = \frac{1}{1-\lambda}$ .

• while  $\lambda$  satisfying last equality and  $\alpha < 2^k \log 2 - k - 5$ ,  $Z = \frac{1}{2}$  is the maximizer and meanwhile,

$$\frac{\mathbf{E}^{2}[U(\psi)]}{\mathbf{E}[U^{2}(\psi)]} \approx \frac{exp(2Nh_{1}(\lambda,\alpha))}{exp(Nh_{2}(\lambda,\alpha,1/2))} = \Omega(1).$$

From discussion above, now we can get the main conclusion in this lecture.

**Theorem 5** The lower bound of  $\alpha$ , in order that a random  $\mathbf{SAT}_N(K, M)$  formula is SAT with vanishing probability in the  $N \to \infty$  limit, can be reached through second moment method as:  $\alpha \geq 2^k \log 2 - k - 5$ .

**Remark 6** Suppose  $\alpha$  such that second moment works and  $\exists Z < \frac{1}{2}$ , s.t.  $h_2(\lambda, \alpha, Z) < 0$ , then there are clusters in the space of solutions. Formally,  $\exists$  pair of solutions at distance  $\frac{N}{2}$ , and  $\nexists$  pair of solutions at distance ZN.