

# Assignment for

## Computer Science Theory for the Information Age

### Day 6

BY ZEN HUANG

5120309027  
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**Exercise 1.** Modify the proof that every monotone property has a threshold for  $G(n, p)$  to apply to the 3-CNF satisfiability problem.

**Answer.**

We'll use  $F(n, m)$  to represent a 3-CNF formula with  $n$  variables,  $m$  clauses.

**Definition 1.** *The increasing property of unsatisfiability.*

*$Q$  is the increasing property of  $F(n, m)$  if when a formula  $F$  has the property any formula obtained by adding clauses to  $F$  must also have the property.*

**Definition 2.**  *$D$ -fold replication of  $F(n, m)$*

*To make an  $M$ -fold replication  $F(n, m')$  of  $F(n, m)$ , first make  $d$  copies of  $F(n, m)$  with clauses generated independently, then include every clause present in any of the  $d$  copies into  $F(n, m')$ .*

**Lemma 3.** *Let  $Q$  be an increasing property of  $F(n, m)$ . for  $m_1 < m_2$*

$$\text{Prob}(F(n, m_1) \text{ has } Q) \leq \text{Prob}(F(n, m_2) \text{ has } Q)$$

**Proof.** *First we generated  $F(n, m_1)$ , then we generate  $F(n, m_2 - m_1)$ , finally we include every clause present in either  $F(n, m_1)$  or  $F(n, m_2 - m_1)$  to get the  $F(n, m_2)$ . By this mean we know that if  $F(n, m_1)$  has property  $Q$   $F(n, m_2)$  must has property  $Q$  as well.  $\square$*

**Theorem.** *Every increasing property  $Q$  of  $F(n, m(n))$  has a threshold at  $m(n) = a$ , where for each  $n$ ,  $a$  is the minimum real number such that*

$$\text{Prob}(F(n, a) \text{ has } Q) = \frac{1}{2}$$

**Proof.**

*Let  $m_0(n)$  be any function such that  $\lim_{n \rightarrow \infty} \frac{m_0(n)}{m(n)} = 0$ . We need to prove that*

$$\lim_{n \rightarrow \infty} \text{Prob}(F(n, m_0(n)) \text{ has } Q) = 0 \tag{1}$$

*Assume that  $\text{Prob}(F(n, m_0(n)) \text{ has } Q)$  does not converge to zero as  $n$  goes to infinity. Then there must be a infinity subsequence  $I$  of  $n$  such that*

$$\exists \varepsilon > 0 \forall n \in I \text{ Prob}(F(n, m_0(n)) \text{ has } Q) > \varepsilon \tag{2}$$

Let  $d = \lceil \frac{1}{\varepsilon} \rceil$ , and let  $F(n, m'(n))$  be the  $d$ -fold replication of  $F(n, m_0(n))$ . Similar proof to Lemma 3 we have

$$\text{Prob}(F(n, m'(n)) \text{ does not have } Q) \leq (\text{Prob}(F(n, m_0(n)) \text{ does not have } Q))^d \quad (3)$$

while

$$m'(n) = 1 - (1 - m_0(n))^d \leq d m_0(n) \quad (4)$$

so that

$$\begin{aligned} \text{Prob}(F(n, d m_0(n)) \text{ does not have } Q) &\leq (\text{Prob}(F(n, m_0(n)) \text{ does not have } Q))^d \\ &\leq (1 - \varepsilon)^d \\ &\leq \frac{1}{e} \\ &\leq \frac{1}{2} \end{aligned} \quad (5)$$

For infinity many  $n$ , since  $m(n)$  is the minimum that  $\text{Prob}(F(n, m(n)) \text{ has } Q) = \frac{1}{2}$ , by Lemma 3,  $d m_0(n) \geq m(n)$ , which contradicts the hypothesis that  $\lim_{n \rightarrow \infty} \frac{m_0(n)}{m(n)} = 0$ .  $\square$

**Exercise 2.** Verify that the sum of  $r$  rank-one matrices  $\sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^T$  can be written as  $\mathbf{U} \mathbf{D} \mathbf{V}^T$ , where the  $\mathbf{u}_i$  are the columns of  $\mathbf{U}$  and  $\mathbf{v}_i$  are the columns of  $\mathbf{V}$ . To do this, first verify that for any two matrices  $\mathbf{P}$  and  $\mathbf{Q}$ , we have

$$\mathbf{P} \mathbf{Q}^T = \sum_i \mathbf{p}_i \mathbf{q}_i^T$$

where  $\mathbf{p}_i$  is the  $i^{\text{th}}$  column of  $\mathbf{P}$  and  $\mathbf{q}_i$  is the  $i^{\text{th}}$  column of  $\mathbf{Q}$ .

**Answer.**

First we prove  $\mathbf{P} \mathbf{Q}^T = \sum_i \mathbf{p}_i \mathbf{q}_i^T$ .

**Proof.**

Let  $\mathbf{A} = \mathbf{P} \mathbf{Q}^T$  and  $\mathbf{B} = \sum_i \mathbf{p}_i \mathbf{q}_i^T$ . Then

$$a_{ij} = \sum_{k=1}^r p_{ik} q_{jk} \quad (6)$$

and

$$\begin{aligned} b_{ij} &= \sum_{i=1}^r (\mathbf{p}_i \mathbf{q}_i^T)_{ij} \\ &= \sum_{k=1}^r p_{ik} q_{jk} \end{aligned} \quad (7)$$

so we have

$$a_{ij} = b_{ij} \quad (8)$$

which is

$$\mathbf{P} \mathbf{Q}^T = \sum_i \mathbf{p}_i \mathbf{q}_i^T \quad (9)$$

□

Then we prove  $\sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^T = \mathbf{U} \mathbf{D} \mathbf{V}^T$ .

**Proof.**

Let  $\mathbf{A} = \mathbf{U} \mathbf{D} \mathbf{V}^T$  and  $\mathbf{B} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^T$ . Then

$$\begin{aligned} b_{ij} &= \sum_{i=1}^r (\sigma_i \mathbf{u}_i \mathbf{v}_i^T)_{ij} \\ &= \sigma_i \sum_{k=1}^r u_{ik} v_{jk} \end{aligned}$$

so if we let  $\mathbf{D}$  be a diagonal matrix with

$$d_{ii} = \sigma_i \tag{10}$$

we can find that

$$a_{ij} = \sum_{k=1}^r u_{ik} \sigma_i v_{jk} \tag{11}$$

which is

$$\mathbf{A} = \mathbf{B} \tag{12}$$

□