

## Lecture 15: Mar 23, 2015

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## 15.1 Lower Bounds for Depth-3 homogeneous circuits

The theme of this lecture is to prove that any depth three homogeneous circuit computing the  $2d^{\text{th}}$  elementary symmetric polynomial in  $n$  variables must have size at least  $(\frac{n}{4d})^d$  over fields of characteristic zero.

**Remark :**

- 1) When  $d = n/c$  (for some constant  $c$ ), we obtain an  $O(c^n)$  exponential lower bound.
- 2) It is clear that symmetric polynomial in  $n$  variables must have degree  $d$  less than  $n$ . However the result does not apply for polynomials of degree  $d$  much greater than  $n$ .

## 15.2 Partial Derivative Method

Let  $f$  be a polynomial. We define the partial derivative measure by

$$PD(f) = \dim[\text{span}\{\partial f\}]$$

where  $\partial f$  is the set of all partial derivatives of  $f$ . In other words we write,

$$PD(f) = \dim[\text{span}\{\partial_S f : S \subseteq [n]\}]$$

where  $\partial_S f$  is the partial derivative  $\frac{\partial f}{\partial x_1 \cdots \partial x_k}$  such that  $S = \{x_1, \dots, x_k\}$ .

### 15.2.1 Properties of Partial Derivatives

Any two polynomials  $f$  and  $g$  over field  $\mathbb{F}$  holds the following properties.

1. Subadditivity :  $PD(f + g) \leq PD(f) + PD(g)$ .
2. Summultiplicativity :  $PD(f \cdot g) \leq PD(f) + PD(g)$ .
3.  $PD(\alpha \cdot f) = \alpha \cdot PD(f)$ , for any  $\alpha \in \{\mathbb{F} \setminus 0\}$

## 15.3 Homogeneous Depth three circuits

Consider a homogeneous circuit  $C$  computing a homogeneous polynomial of degree  $d$  in  $n$  variables  $\{x_1, x_2, \dots, x_n\}$ . Let  $C$  be

$$C = T_1 + T_2 + \cdots + T_s$$

such that  $T_i = \prod_{j=1}^d l_{ij}$ , where  $l_{ij}$  are linear forms in variables  $x_1, x_2, \dots, x_n$ .

**Lemma 15.1**  $PD(C) \leq s \cdot 2^d$

**Proof:**

$$\begin{aligned}
 PD(C) &\leq \sum_{i=1}^s PD(T_i) && \text{(using subadditivity)} \\
 &\leq \sum_{i=1}^s \prod_{j=1}^d PD(l_{ij}) && \text{(using submultiplicativity)} \\
 &\leq \sum_{i=1}^s 2^d \\
 &\leq s \cdot 2^d
 \end{aligned}$$

**Notation :** Let  $S_n^{2d}$  denote the  $2d^{th}$  degree elementary symmetric polynomial in  $n$  variables.

**Theorem 15.2**  $PD(S_n^{2d}) \geq \binom{n}{d}$  over fields of characteristic zero.

Before proving the theorem, let us prove the required lower bound. From *theorem 15.2* and *lemma 15.1*, we get,

$$\begin{aligned}
 \binom{n}{d} &\leq s \cdot 2^{2d}, \quad \text{if } C \text{ computes the polynomial } S_n^{2d} \\
 \implies s &\geq \frac{\binom{n}{d}}{2^{2d}} \\
 &\geq \left(\frac{n}{4d}\right)^d && \text{(using Stirling formula)}
 \end{aligned}$$

Hence we have proved the required lower bound. Now the remainder of the proof is to prove *theorem 15.2*.

### 15.3.1 Proof of Theorem 15.2

Here we restrict our focus only to partial derivatives of order  $d$ . Let  $T \subseteq \{x_1, \dots, x_n\}$  and  $|T| = d$ . Now,

$$\partial_T S_n^{2d} = \sum_{W \subseteq [n] \text{ \& } |W|=d \text{ \& } W \cap T = \emptyset} \prod_{i \in W} x_i \tag{15.1}$$

Let us define a column vector  $\mathbf{m}_{\binom{n}{d} \times 1}$ , whose rows are indexed by subsets of  $[n]$  of size  $d$ . We define the entries by  $\partial_T S_n^{2d}$  for any row identified by  $T$ . That is

$$\mathbf{m}_{\binom{n}{d} \times 1} = [\partial_T S_n^{2d}] \tag{15.2}$$

From (15.1) and (15.2), we write

$$\mathbf{m} = D \cdot \mathbf{v}$$

where,  $D_{\binom{n}{d} \times \binom{n}{d}}$  is a 0/1 matrix whose rows and columns are identified by subsets of  $[n]$  of size  $d$  such that  $D_{T,W} = 1$  iff  $T \cap W = \emptyset$ . Here,  $\mathbf{v}_{\binom{n}{d} \times 1}$  is a column vector whose rows are identified similar to columns of  $D$  such that  $\mathbf{v}_W = \prod_{i \in W} x_i$ .

From now on let us call  $D$  as the disjoint matrix.

**Theorem 15.3** *The disjoint matrix  $D$  has maximal rank over any field of characteristic zero. That is,  $\text{rank}(D) = \binom{n}{d}$ .*

Since  $D$  has maximal rank, we can say that  $d^{\text{th}}$  order partial derivatives of  $S_n^{2d}$ , that is, the entries of  $\mathbf{m}$  are all linearly independent. Therefore, we obtain the result  $PD(S_n^{2d}) \geq \binom{n}{d}$ . Now let us prove *theorem 15.3*.

## 15.4 Proof of Theorem 15.3

Let  $S = \{1, 2, \dots, n\}$ . Let us call any subset of  $S$  of size  $l$  and  $k$  as  $l$ -set and  $k$ -set respectively. For any two positive integers  $l, k$  ( $k \leq l$ ) we construct a 0/1 incidence matrix  $B$  whose rows and columns are indexed by the set of all possible  $l$ -sets and  $k$ -sets respectively such that,  $B_{ij} = 1$ , iff the  $l$ -set corresponding to the  $i^{\text{th}}$  row contains the  $k$ -set corresponding to the  $j^{\text{th}}$  column. Clearly  $B$  has  $\binom{n}{l}$  rows and  $\binom{n}{k}$  columns. It has been proved by *Gottlieb* [2] that the rank of such matrix  $B$  is maximal. That is  $\text{rank}(B) = \min\{\binom{n}{l}, \binom{n}{k}\}$ .

The idea here is to show a reduction from disjoint matrix  $D$  to matrix  $B$  and hence claiming that  $\text{rank}(D)$  is  $\binom{n}{d}$ . The reduction is as follows.

Suppose  $d \leq n/2$ . Here the  $i^{\text{th}}$  row is identified by the set  $T_i$  of size  $d$ . Also the  $j^{\text{th}}$  column is identified by a set with  $d$  elements (say  $W_j$ ). We can also identify the same column  $j$  by the set  $[n] \setminus W_j$ . Clearly we observe that,  $T_i$  and  $W_j$  are disjoint if and only if  $T_i$  is contained in the set  $[n] \setminus W_j$ . Therefore, we claim that the disjoint matrix  $D$  is same as an incidence matrix  $B$ . Hence  $D$  has maximal rank. That is,  $\text{rank}(D) = \binom{n}{d}$ .

Similarly we can prove the case when  $d > n/2$ .

## 15.5 The incidence matrix $B$ has maximal rank<sup>[2]</sup>

We recall the 0/1 matrix  $B$  whose rows and columns are indexed by set of all possible  $l$ -sets and  $k$ -sets respectively. The  $(i, j)$  element of matrix  $B$  takes value 1, only if the  $l$ -set corresponding to the  $i^{\text{th}}$  row contains the  $k$ -set corresponding to the  $j^{\text{th}}$  column.

Let us define lexicographical ordering for  $m$ -sets and  $n$ -sets. We represent any  $m$ -set by the vector  $(a_1, a_2, \dots, a_m)$ , where  $a_i < a_{i+1}$ , for  $i \in [m-1]$ . Also we say  $(a_1, a_2, \dots, a_m) < (b_1, b_2, \dots, b_m)$ , if and only if  $a_i < b_i$  for the smallest value of  $i$  such that  $a_i \neq b_i$ . Let us call this ordering the canonical ordering.

Now we define the canonical matrix  $A_{l,k}^n$  obtained by ordering the rows and columns of  $B$  in their canonical order. Clearly  $\text{rank}(A_{l,k}^n) = \text{rank}(B)$ . Therefore, it is sufficient to prove that rank of  $A_{l,k}^n$  is maximal.

**Notations :** We use the notation  $R_{l,k}^n$ ,  $C_{l,k}^n$  to represent the row null space and column null space of the matrix  $A_{l,k}^n$  respectively.

**Lemma 15.4** *It is easy to observe that the matrix  $A_{l,k}^n$  has,*

1.  $\binom{n}{l}$  rows
2.  $\binom{n}{k}$  columns
3.  $\binom{l}{k}$  1's in each row
4.  $\binom{n-l}{l-k}$  1's in each column

**Lemma 15.5** *We can also verify the following*

1.  $A_{1,1}^1 = A_{1,0}^1 = A_{0,0}^1 = A_{0,0}^0 = [1]$
2.  $A_{l,l}^n = I_l^n$ , where  $I_l^n$  is the  $\binom{n}{l} \times \binom{n}{l}$  identity matrix.

**Lemma 15.6** *By definition of  $A_{l,k}^n$ , we represent  $A_{l,k}^n$  by the following partition formula.*

$$A_{l,k}^n = \begin{bmatrix} A_{l-1,k-1}^{n-1} & A_{l-1,k}^{n-1} \\ O & A_{l,k}^{n-1} \end{bmatrix}_{\binom{n}{l} \times \binom{n}{k}}$$

**Lemma 15.7** *For  $n \geq l \geq p \geq k \geq 0$ ,*

$$A_{l,p}^n \cdot A_{p,k}^n = \binom{l-k}{p-k} A_{l,k}^n$$

**Proof sketch :** The above formula is proved by induction on  $n$ . Clearly the base case ( $n = 1$ ) can be verified (using lemma 15.5). The induction step uses the partition formula (lemma 15.6) thus facilitating the multiplication of  $A_{l,p}^n$  and  $A_{p,k}^n$ , which results in  $\binom{l-k}{p-k} A_{l,k}^n$ .

**Theorem 15.8**  $\dim(R_{l-1,k}^{n-1}) + \dim(R_{l,k-1}^{n-1}) = \dim(R_{l,k}^n)$ .

**Proof Sketch :** Let us define a matrix  $T = \begin{bmatrix} I_{l-1} & 0 \\ \frac{-1}{l-k} A_{l,l-1}^{n-1} & I_m \end{bmatrix}_{\binom{n}{l} \times \binom{n}{l}}$ .

On premultiplying  $A_{l,k}^n$  by  $T$  we get,

$$T.A_{l,k}^n = \begin{bmatrix} A_{l-1,k-1}^{n-1} & A_{l-1,k}^{n-1} \\ \frac{-(l-k+1)}{l-k} A_{l,k-1}^{n-1} & 0 \end{bmatrix}_{\binom{n}{l} \times \binom{n}{k}}$$

Consider a vector  $\mathbf{v} = (\mathbf{x}, \mathbf{y})$  such that

$$\mathbf{v}.T.A_{l,k}^n = (\mathbf{x}.A_{l-1,k-1}^{n-1} - \frac{(l-k+1)}{(l-k)}\mathbf{y}.A_{l,k-1}^{n-1}, \mathbf{x}.A_{l-1,k}^{n-1})$$

On solving  $\mathbf{v}.T.A_{l,k}^n = 0$ , we get  $\mathbf{x}.A_{l-1,k}^{n-1} = 0$  and  $\mathbf{y}.A_{l,k-1}^{n-1} = 0$ . (uses lemma 15.7 and the fact that field  $\mathbb{F}$  has characteristic zero).

Also since  $T$  is a non-singular matrix, we obtain that  $\mathbf{v}$  is a direct sum of  $\mathbf{x}$  and  $\mathbf{y}$ , implying

$$\dim(R_{l-1,k}^{n-1}) + \dim(R_{l,k-1}^{n-1}) = \dim(R_{l,k}^n).$$

**Theorem 15.9**  $\dim(C_{l-1,k}^{n-1}) + \dim(C_{l,k-1}^{n-1}) = \dim(C_{l,k}^n)$ .  
(We can prove the theorem similar to theorem 15.8).

**Corollary 15.10** .

$$\begin{aligned} 1. \dim(R_{l,k}^n) &= \begin{cases} 0 & \text{if } l+k > n, \\ \binom{n}{l} - \binom{n}{k} & \text{if } l+k \leq n. \end{cases} \\ 2. \dim(C_{l,k}^n) &= \begin{cases} \binom{n}{k} - \binom{n}{l} & \text{if } l+k > n, \\ 0 & \text{if } l+k \leq n. \end{cases} \end{aligned}$$

**Proof Sketch :** We prove this by induction on  $n$ . We can easily verify the base case when  $n = 1$ . We analyze the induction by two cases. Suppose  $l+k > n$ . This implies  $(l-1)+k > n-1$  and  $l+(k-1) > n-1$ . By induction hypothesis, we get  $\dim(R_{l-1,k}^{n-1}) = \dim(R_{l,k-1}^{n-1}) = 0$ . Using theorem 15.8 we get,  $\dim(R_{l,k}^n) = 0$ . Also by rank-nullity theorem we obtain  $\dim(C_{l,k}^n) = \binom{n}{k} - \binom{n}{l}$ . Therefore (by rank-nullity theorem) we get, the rank of the matrix is equal to the number of rows, that is  $\text{rank}(A_{l,k}^n) = \binom{n}{l}$ .

Similarly when  $l+k \leq n$ , we get,  $\text{rank}(A_{l,k}^n) = \binom{n}{k}$ . Thus we conclude by stating that the rank of matrix  $A_{l,k}^n$  is maximal.

## 15.6 Motivation for homogeneous circuits

Consider the polynomial

$$\begin{aligned} f(z, x_1, x_2, \dots, x_n) &= (z + x_1) \cdot (z + x_2) \cdots (z + x_n) \\ &= c_n z^n + c_{n-1} z^{n-1} + \cdots + c_0 \end{aligned}$$

where the coefficient of  $z^{n-d}$ ,  $c_{n-d}$  is exactly the elementary symmetric polynomial  $S_n^d(x_1, x_2, \dots, x_n)$ . Let the evaluations of  $f(z, x_1, x_2, \dots, x_n)$  at  $z = \alpha_1, \alpha_2, \dots, \alpha_{n+1}$  be  $g_1(\mathbf{x}), g_2(\mathbf{x}), \dots, g_{n+1}(\mathbf{x})$ , where  $\alpha_1, \alpha_2, \dots, \alpha_{n+1}$  are distinct field elements. That is, for all  $i \in [n+1]$ ,

$$\begin{aligned} f(\alpha_i, x_1, x_2, \dots, x_n) &= g_i(\mathbf{x}) \\ &= c_n \alpha_i^n + c_{n-1} \alpha_i^{n-1} + \cdots + c_0 \end{aligned}$$

In matrix notation we write,

$$\mathbf{g} = A\mathbf{c}$$

$$\text{where, } \mathbf{g} = \begin{bmatrix} g_1(\mathbf{x}) \\ g_2(\mathbf{x}) \\ \vdots \\ g_{n+1}(\mathbf{x}) \end{bmatrix}_{(n+1) \times 1}, \quad A = \begin{bmatrix} \alpha_1^0 & \alpha_1^1 & \alpha_1^2 & \cdots & \alpha_1^n \\ \alpha_2^0 & \alpha_2^1 & \alpha_2^2 & \cdots & \alpha_2^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \alpha_{n+1}^0 & \alpha_{n+1}^1 & \alpha_{n+1}^2 & \cdots & \alpha_{n+1}^n \end{bmatrix}_{(n+1) \times (n+1)}, \quad \mathbf{c} = \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_n \end{bmatrix}_{(n+1) \times 1}$$

Since  $A$  is a Vandermonde matrix, we know that inverse of  $A$  exists. Therefore the coefficient vector  $\mathbf{c}$  can be computed by,

$$\mathbf{c} = A^{-1} \cdot \mathbf{g}$$

That is, for  $0 \leq j \leq n$ ,

$$c_j = \sum_{k=1}^{n-1} \beta_k \cdot g_k(\mathbf{x})$$

where,  $\beta_k \in \mathbb{F}$  is a field constant. Clearly  $c_j$  can be computed by a depth three circuit (not homogeneous) of polynomial size. That is, there exists a depth three circuit of polynomial size that computes the elementary symmetric polynomial. Thus we state the following corollary.

**Corollary 15.11** *We cannot homogenize a depth-three circuit without a superpolynomial loss in size.*

## References

- [1] NOAM NISAN and AVI WIGDERSON, Lower Bounds on Arithmetic Circuits via Partial Derivatives, *Computational Complexity*, 1997
- [2] D.H. GOTTLIEB, A certain class of Incidence Matrices, *Proceedings of the American Mathematical Society*, Volume 17, Issue 6, Dec., 1966