

## Communication Complexity of Matrix Computation over Finite Fields\*

J. I. Chu<sup>1</sup> and G. Schnitger<sup>2</sup>

<sup>1</sup>Computer Science Department, New Mexico State University,  
Las Cruces, NM 88003, USA

<sup>2</sup>Computer Science Department, Pennsylvania State University,  
University Park, PA 16802, USA

**Abstract.** We investigate the communication complexity of *singularity testing* in a finite field, where the problem is to determine whether a given square matrix  $M$  is singular. We show that, for  $n \times n$  matrices whose entries are elements of a finite field of size  $p$ , the communication complexity of this problem is  $\Theta(n^2 \log p)$ . Our results imply tight bounds for several other problems like *determining the rank* and *computing the determinant*.

### 1. Introduction

Parallel computing saves time by spreading the work load among a number of active elements. However, this may create the necessity of communication among those active elements. Yao [12], [13] introduced the deterministic model of *communication complexity* to measure this additional computational resource. In this model there are two agents performing the desired computation cooperatively. The agents have their own memory and can exchange messages with each other. When the computation starts, each agent receives one-half of the input bits according to a fixed partition rule  $\pi$ . Then they execute a fixed protocol  $P$  which leads the two agents to perform local computations and communicate messages according to their own information and previously obtained messages. At the end of the computation, each agent knows the values of the output variables for which it is responsible. The number of bits exchanged represents the communication requirement for this particular computation.

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Let  $f$  be the function to be computed by a parallel system and let  $N$  be the number of input bits. Call any partition rule which divides these  $N$  bits into two even portions an even partition. Every even partition,  $\pi$ , together with a correct protocol,  $P$ , represents a possible design of a system computing  $f$ . Let  $C(\pi, P)$  be the maximum number of bits to be exchanged for some input instance of size  $N$  under partition  $\pi$  and protocol  $P$ . The communication complexity of  $f$  under partition  $\pi$ ,  $C(\pi)$ , is the minimum of  $C(\pi, P)$  over all correct protocols  $P$  for  $\pi$ . Then the communication complexity of  $f$  is defined to be the minimum of  $C(\pi)$  over all even partitions  $\pi$ .

Thompson's [9] area-time tradeoffs can be best expressed by communication complexity. Suppose  $f$  has  $N$  input bits and a VLSI chip computing  $f$  has area  $A$  and worst-case computation time  $T$ . Thompson shows the tradeoff  $AT^2 = \Omega(f^2)$ . Subsequently, the area bound  $A = \Omega(f)$  was proven [2], [10], [13]. Combining these two formulas, we get  $AT^{2\alpha} = \Omega(f^{1+\alpha})$ ,  $0 \leq \alpha \leq 1$ .

We investigate the communication complexity of testing whether a matrix over a finite field is singular.

**Theorem 1.1.** *Let  $F_p$  be a finite field with  $p$  elements. Let  $M$  be a square matrix of dimension  $n$ , where each entry consists of  $\lceil \log p \rceil$  bits encoding an element of  $F_p$ . (All logarithms in this paper are base 2.) The communication complexity of "deciding whether  $M$  is nonsingular" is  $\Theta(n^2 \log p)$ .*

Chu and Schnitger [4] showed that when entries of  $M$  are integers of up to  $k$  bit long, the corresponding communication complexity is  $\Theta(kn^2)$ . The proof for the integer case relies on the ability of encoding vectors of very large components by matrices. That is not applicable in a finite field as any value can be represented by  $\log p$  bits.

The commonly used *transitivity* approach of Vuillemin [10] does not seem to work for our problem, either. Vuillemin's approach is successful for many functions or languages that are powerful enough to express the *identity* problem (given two strings  $x$  and  $y$ , are  $x$  and  $y$  identical?). However, there does not seem to be a large enough identity problem embedded in the singularity testing problem.

We utilize the following property of vector spaces over a finite field to bound the size of monochromatic submatrices in the truth matrix of our problem (see Section 2). Let  $F_p$  be a finite field of  $p$  elements. Assume  $S_1$  and  $S_2$  are two sets of dimension  $m$  subspaces in  $F_p^{2m}$  such that, for any  $A \in S_1$  and any  $B \in S_2$ ,  $A \cap B$  contains only the zero vector. Then  $\# S_1 \cdot \# S_2 \leq p^{m^2+2m}$  (Lemma 3.7).

We prove Theorem 1.1 in Section 3. Theorem 1.1 also establishes the communication complexity of the following problems.

**Corollary 1.2.** *Let  $M$  be defined as in Theorem 1.1. The communication complexity of the following problems is  $\Theta(n^2 \log p)$ .*

- (a) Computing the determinant of  $M$ .
- (b) Computing the rank of  $M$ .

Roundoff errors are inevitable when computing in the ring of real numbers using digital computers. In applications where Fourier transform is used to compute a convolution efficiently, an exact result is often required. However, the fast Fourier transform (FFT) implementation of cyclic convolution introduces significant amounts of roundoff errors [6]. These errors can be avoided by computing over a suitable finite field. Agarwal and Burrus [1] give efficient implementation of finite transforms over rings of integers modulo Fermat numbers. They report a much faster performance over the best FFT implementation. Convolution over a finite field also speeds up multiplications of long integers [7]. Another important potential of computing in finite fields is the use of residue number systems in arithmetic computations [8]. VLSI implementation computing complex inner products using residue arithmetic has been proposed [11].

Let  $X$  be a finite set of vectors from a vector space  $U$  and let  $L$  be the set of subspaces  $\{V: V \text{ is spanned by some subset of } X\}$ . The *Vector Space Span* problem is defined as follows: Given two elements of  $L$ ,  $V_1$  and  $V_2$ , decide whether their union spans  $U$ .

Lovász and Saks [5] determined the *fixed-partition* communication complexity of the vector space span problem to be  $\Theta(\log(\#L))$ , where one agent reads  $V_1$  and the other reads  $V_2$ . Theorem 1.1 also establishes the *unrestricted* communication complexity of this problem when  $X$  is chosen to be the set of vectors over some finite field.

**Corollary 1.3.** *When  $X = \mathbb{F}_p^n$ , the communication complexity of the vector space span problem is  $\Theta(n^2 \log p)$ .*

We present a sketch of our proof techniques in Section 2. The proof of Theorem 1.1 is given in Section 3.

## 2. Proof Techniques

Let  $f$  be some function to be computed using the communication complexity model. Let us fix the size of the input to be  $N$  bits and the input partition to be  $\pi$ . Then the computation can be viewed as computing a function of two arguments, in which the first argument consists of the  $N/2$  input bits given to the first agent and the second argument consists of the remaining bits.

In case the output consists of one bit, we can characterize the computation of a two-argument boolean function by a *Truth Matrix*. Each possible instance of the first argument takes up one row of the truth matrix and each possible instance of the second argument takes up one column. The matrix entry at the intersection of a particular pair of row and column contains the output bit corresponding to the arguments assigned to this row and this column.

A submatrix of the truth matrix is a *monochromatic submatrix* if it contains only a single value. More specifically, if this value is 1, the monochromatic submatrix is called 1-chromatic; otherwise it is called 0-chromatic. Yao [12] shows that, given the input partition  $\pi$  (and hence the truth matrix) of some boolean-valued

function  $f$ , the communication complexity of  $f$  under partition  $\pi$  is at least  $\log d(f) - 2$ , where  $d(f)$  is the minimal number of disjoint monochromatic submatrices that partition the truth matrix of  $f$ .

Suppose when the input dimension,  $n$ , of the singularity testing is odd, the communication complexity is  $\alpha(n^2 \log p)$ . Then the  $\alpha(n^2 \log p)$  bound also holds for the case when  $n$  is even. This is because, for any even  $n$ , the agents can pretend that there are one additional row and one additional column in the input matrix. All the entries on the  $(n + 1)$ st row and the  $(n + 1)$ st column are 0 except that the  $[n + 1, n + 1]$  entry is nonzero. Therefore, we only consider even input dimensions when proving Theorem 1.1.

For the proof of Theorem 1.1, we first tackle the case of a particular input partition as defined below; then we show that arbitrary partitions do not change the communication complexity asymptotically.

**Definition 2.1.** Assume the input is a  $2m \times 2m$  matrix. Let  $\pi_0$  denote the following input partition: the first agent receives all the bits encoding the entries in the first  $m$  columns; the second agent receives the other half.

Given  $\pi_0$ , we define an entry of the truth matrix to be “one” if the corresponding input matrix is nonsingular. We show that there is a large submatrix,  $T$ , of the truth matrix such that:

- (1)  $T$  contains a large number ( $>p^{n^2/2}$ ) of “one” entries.
- (2) Every 1-chromatic submatrix of  $T$  is of relatively small size ( $\leq p^{n^2/4+n}$ ).

This allows us to apply Yao’s lower bound method.

The first property of  $T$  as given above is easy to prove. We use induction to establish the second property. The inductive step involves proving a  $\Theta(m \log p)$  bound on the communication complexity of “the inner product over  $F_p$ .” In this new problem each agent reads a vector in  $F_p^m$  and tries to decide whether the two input vectors have a nonzero inner product (“one” entries in its truth matrix represent nonzero products). In this proof we use techniques of Chor and Goldreich [3]. In particular, we utilize that the truth matrix of the inner product function can be made orthogonal if we replace the value 0 by  $1 - p$ .

### 3. Singularity Testing over Finite Fields

In this section we prove Theorem 1.1 Let  $F_p$  be a finite field with  $p$  elements. Our input, an  $n \times n$  matrix over  $F_p$ , can be represented by  $n^2 \log p$  bits. First, we establish the  $\Theta(n^2 \log p)$  bound on the communication complexity of this problem under the assumption that the input is partitioned according to  $\pi_0$  (see Definition 2.1). Note that the upper bound is trivial.

We restrict our attention to a submatrix  $T$  of the truth matrix. Interpret each column of the input matrix as a vector in  $F_p^n$ . Then each row or column in the truth matrix corresponds to a subspace spanned by  $n/2$  vectors. We select for  $T$  exactly one row and one column that correspond to each dimension  $n/2$  subspaces of  $F_p^n$ .

The entry  $[i, j]$  in  $T$  is set to 1 whenever the two subspaces corresponding to the  $i$ th row and  $j$ th column intersect only at the zero vector, i.e., whenever the input matrix is nonsingular.

**Lemma 3.1.** *The number of “one” entries in  $T$  is at least  $p^{n^2/2}$ .*

*Proof.* First we count the number of “one” entries in a row of  $T$ . Assume that a row is labeled by a vector space  $W$  of dimension  $n/2$ . Consider the number of distinct bases which span subspaces of dimension  $n/2$  and intersect  $W$  only at the zero vector. Let us construct such bases by picking  $n/2$  base vectors one by one.

If  $V \subset F_p^n$  is a subspace of dimension  $d$ , there are exactly  $p^n - p^d$  vectors outside of  $V$ . We have  $\#(F_p^n \setminus W) = p^n - p^{n/2}$  choices for the first base vector and  $p^n - p^{n/2+1}$  choices for the second base vector after the first one is chosen. In general, once the first  $i$  base vectors are chosen we have  $p^n - p^{n/2+i}$  choices for the  $(i + 1)$ st base vector. So the total number of bases is  $\prod_{i=n/2}^{n-1} (p^n - p^i)$ .

The same counting argument shows that there are  $\prod_{j=0}^{n/2-1} (p^{n/2} - p^j)$  distinct bases for each subspace of dimension  $n/2$ . Therefore, the number of “one” entries in any row of  $T$  is equal to

$$\prod_{j=0}^{n/2-1} \frac{p^n - p^{n/2+j}}{p^{n/2} - p^j} = \prod_{j=0}^{n/2-1} p^{n/2} = p^{n^2/4}.$$

Since  $T$  is symmetric, there are  $p^{n^2/4}$  “one” entries in each column of  $T$ . This means  $T$  has at least  $p^{n^2/4}$  rows. Therefore, the number of “one” entries in  $T$  is at least  $p^{n^2/2}$ . □

Let  $IPZ(m, p)$  denote the problem of deciding whether the inner product of two vectors in  $F_p^m$  is nonzero. Let  $\pi_1$  be the input partition of  $IPZ(m, p)$  such that each agent reads one of the input vectors. We now establish an intermediate result which states that  $IPZ(m, p)$  has communication complexity  $\Theta(m \log p)$  under partition  $\pi_1$ . This generalizes a result of Chor and Goldreich [3] for the inner product over  $F_2$ .

Let  $\Psi$  be the set of dimension  $m - 1$  subspaces of  $F_p^m$ .  $\Psi$  contains  $\#\Psi = (p^m - 1)/(p - 1)$  elements because a hyperplane is specified by one linear equation which is unique up to the  $p - 1$  multiples of its coefficients. We set up an auxiliary matrix  $Z$  as follows. Create one row for each space in  $\Psi$  and one column for each vector in  $F_p^m$ . For the entry at row  $Q$  and column  $v$ , set its value to  $(1 - p)$  if  $v \in Q$ ; otherwise, set it to 1.

A vector  $v$  is orthogonal to a subspace  $Q$  if and only if the inner product  $v \cdot t$  is 0 for all  $t \in Q$ . Each row of  $Z$  also corresponds to the  $p$  vectors orthogonal to  $Q$ . Another way to interpret  $Z$  is that each row or column of  $Z$  corresponds to a vector; an entry is  $(1 - p)$  if and only if the two vectors have inner product 0. Hence,  $Z$  is also a submatrix of the truth matrix of  $IPZ(m, p)$  under  $\pi_1$  if we replace the value  $(1 - p)$  by 0.

**Lemma 3.2.** *The rows of  $Z$  are pairwise orthogonal.*

*Proof.* We use  $r_i$  to denote the  $i$ th row vector of  $Z$ . Pick any two different row vectors of  $Z$ , say  $r_i$  and  $r_j$ . Let  $Q_i$  and  $Q_j$  be the elements of  $\Psi$  which correspond to these two row vectors, respectively. Then  $r_i \cdot r_j = \sum_{k=1}^m (Z[i, k] \cdot Z[j, k])$ . Let  $v$  be the vector corresponding to the  $k$ th column of  $Z$ . Then

$$Z[i, k] \cdot Z[j, k] = \begin{cases} (1-p)^2, & \text{if } v \in (Q_i \cap Q_j), \\ 1, & \text{if } v \notin (Q_i \cup Q_j), \\ 1-p, & \text{otherwise.} \end{cases}$$

Note that  $\#(Q_i \cap Q_j) = p^{m-2}$  and  $\#(Q_i \cup Q_j) = 2p^{m-1} - p^{m-2}$ . We obtain  $r_i \cdot r_j = (p-1)^2 p^{m-2} + (p^m - 2p^{m-1} + p^{m-2}) - (p-1)(2p^{m-1} - 2p^{m-2}) = p^{m-2}(2(p-1)^2 - 2(p-1)^2) = 0$ . □

**Lemma 3.3.** *The size of any 1-chromatic submatrix of  $Z$  cannot exceed  $(p-1)p^m$ .*

*Proof.* Pick any  $a$  rows and  $b$  columns of  $Z$ . Since we do not demand any ordering of rows and columns, we assume the first  $a$  rows and the first  $b$  columns are chosen in order to simplify notation. Define  $x = \sum_{i=1}^a \sum_{j=1}^b Z[i, j]$ . If the submatrix defined by these  $a$  rows and  $b$  columns is 1-chromatic, then  $x^2$  must equal  $(a \ b)^2$ . It suffices to prove that  $x^2 \leq ab(p-1)p^m$ .

Observe from the proof of Lemma 3.2 that, for every  $i$ ,  $r_i \cdot r_i = (p-1)^2 p^{m-1} + (p^m - p^{m-1}) = (p-1)p^{m-1}(p-1+1) = (p-1)p^m$ . Recall Cauchy's inequality:  $(v \cdot u)^2 \leq \|v\|^2 \cdot \|u\|^2$ .

$$\begin{aligned} x^2 &= \left( \sum_{j=1}^b \left( 1 \cdot \sum_{i=1}^a Z[i, j] \right) \right)^2 && \text{interpret the outer summation} \\ &&& \text{as inner product and apply} \\ &&& \text{Cauchy's inequality} \\ &\leq b \sum_{j=1}^b \left( \sum_{i=1}^a Z[i, j] \right)^2 \\ &\leq b \sum_{j=1}^b \left( \sum_{i=1}^a Z[i, j] \right)^2 \\ &= b \left( \sum_{i=1}^a r_i \cdot \sum_{i=1}^a r_i \right) && \text{apply orthogonality} \\ &= b \sum_{i=1}^a (r_i \cdot r_i) \\ &= ab(p-1)p^m. \end{aligned}$$

**Theorem 3.4.** *The communication complexity of  $IPZ(m, p)$  under partition  $\pi_1$  is  $\Theta(m \log p)$ .*

*Proof.* Since  $Z$  has  $\#\Psi = (p^m - 1)/(p - 1)$  rows and each rows has  $p^m - p^{m-1}$  “one” entries, there are  $(p^m - 1)p^{m-1}$  “one” entries in  $Z$ . Applying Yao’s lower-bound method, we obtain that the communication complexity of  $IPZ(m, p)$  under  $\pi_1$  is at least  $\Omega(\log((p^m - 1)/p(p - 1))) = \Omega(m \log p)$ . On the other hand, the input read by each agent has  $m \log p$  bits, which implies a trivial upper bound.  $\square$

We restate Lemma 3.3 in a form needed in subsequent proofs.

**Lemma 3.5.** *Let  $Q_1, \dots, Q_t$  be elements of  $\Psi$ . Define  $D$  to be the complement of  $Q_1 \cup \dots \cup Q_t$  with respect to  $F_p^n$ . Then  $t \times (\#D) \leq (p - 1)p^m$ .*

*Proof.* The  $r$  rows of  $Z$  corresponding to  $Q_i, 1 \leq i \leq t$ , and the  $\#D$  columns corresponding to vectors in  $D$  form a 1-chromatic submatrix of  $Z$ .  $\square$

Let  $Y_1$  be any subset of  $F_p^n$  which contains  $\mathbf{0}$  (the zero vector). Let  $Y_2$  denote the set  $(F_p^n \setminus Y_1) \cup \{\mathbf{0}\}$ . Observe that any choice of  $Y_1$  specifies a 1-chromatic submatrix,  $S$ , of  $T$ —the rows (columns) of  $S$  correspond to the dimension  $n/2$  subspaces contained in  $Y_1$  ( $Y_2$ ). Conversely, every 1-chromatic submatrix induces canonically a set  $Y_1$ .

**Definition 3.6.** Let  $i$  be any integer in the range  $[0, n/2]$ . Set  $m := i + n/2$ . Let  $Y_1$  and  $Y_2$  be any two subsets of  $F_p^m$  such that  $\mathbf{0} \in Y_1$  and  $Y_2 = (F_p^m \setminus Y_1) \cup \{\mathbf{0}\}$ . We call any dimension  $n/2$  subspace a  $Y_1$ -space if it is contained in  $Y_1$ . Also, any dimension  $i$  subspace contained in  $Y_2$  is called a  $Y_2$ -space. Let  $s(Y_1)$  and  $t(Y_2)$  be the number of  $Y_1$ -spaces and  $Y_2$ -spaces, respectively. Define  $E(i)$  as the maximum, over all choices of  $Y_1$ , of the product  $s(Y_1) \cdot t(Y_2)$ .

We now investigate the fixed partition case of the singularity testing problem. Note that when  $i = n/2$ , every  $Y_2$ -space corresponds to a column in some 1-chromatic submatrix of  $T$ , and every  $Y_1$ -space corresponds to a row. Hence,  $E(n/2)$  is the maximum size of any 1-chromatic submatrix.

**Lemma 3.7.** *The maximum size of any 1-chromatic submatrix of  $T$  is at most  $p^{n^2/4+n}$ .*

*Proof.* We claim that, for all  $0 \leq i \leq n/2$ ,  $E(i) \leq p^{i \cdot n/2 + 2i}$ . Hence,  $E(n/2) \leq p^{n^2/4+n}$ . This claim is proved by induction on  $i$ .

*Basis:*  $i = 0$ .

We have either  $Y_1 = F_p^{n/2}$  or there are no  $Y_1$ -spaces. In the former case  $s(Y_1) = 1$  and  $t(Y_2) = 1$ . Therefore,  $E(0) \leq 1$ .

*Induction Step:* Assume that  $E(i - 1) \leq p^{(i-1)n/2 + 2i - 2}$  for some  $i \leq n/2$ . We prove that  $E(i) \leq p^{i \cdot n/2 + 2i}$ .

Let  $m, Y_1$ , and  $Y_2$  be defined as in Definition 3.6 such that  $E(i)$  is maximized. Let  $\Psi$  be the set of dimension  $m - 1$  subspaces of  $F_p^m$ . Recall that a  $Y_1$ -space is a dimension  $n/2$  subspace contained in  $Y_1$  and  $Y_2$ -space is a dimension  $i$  subspace contained in  $Y_2$ .

Let  $Q$  be the element of  $\Psi$  which contains the largest number of  $Y_1$  spaces. Let this number be  $r > 0$ . Consider the set of dimension  $i - 1$  spaces contained in  $Y_2 \cap Q$ . Assume there are  $c$  of them. Then  $r \cdot c \leq p^{(i-1)n/2+2i-2}$  by the induction hypothesis. Also, each  $Y_2$ -space intersects  $Q$  at exactly one of these  $c$  spaces. To see this, let  $R_1$  be any  $Y_1$ -space in  $Q$  and let  $K$  be the intersection of any  $Y_2$ -space and  $Q$ . Then  $K$  must be a vector space. Since  $R_1$  and  $K$  only intersect at the zero vector, the dimension of  $K$  is no more than  $i - 1$ . Since  $R_1$  and any  $Y_2$ -space span  $F_p^m$ , the dimension of  $K$  is no less than  $i - 1$ .

Let  $K$  be an arbitrary dimension  $i - 1$  vector space contained in  $Y_2 \cap Q$ . In order to bound  $t(Y_2)$ , the total number of  $Y_2$ -spaces, we first bound the number of  $Y_2$ -spaces containing  $K$ .

We can generate a  $Y_2$ -space containing  $K$  by adding one more base vector,  $\mathbf{b}$ , to the basis of  $K$ . Let  $R_1$  be any  $Y_1$ -space. Then  $\mathbf{b}$  cannot belong to the dimension  $m - 1$  subspace  $R_1 + K$ . Assume that there are  $l$  different spaces  $R_i + K$  (for  $Y_1$ -spaces  $R_1, \dots, R_l$ ). We call  $l$  the *stiffness* of  $K$ . Then, by Lemma 3.5, there are at most  $(p - 1)p^{m/l}$  choices for  $\mathbf{b}$ .

Note that each  $Y_2$ -space thus generated from  $K$  contains  $(p - 1)p^{i-1}$  possible choices of  $\mathbf{b}$ . This is because a  $Y_2$ -space contains  $p^i$  vectors and each of them, except those  $p^{i-1}$  vectors in  $K$ , may be used as  $\mathbf{b}$ . Therefore, the total number of  $Y_2$ -spaces containing  $K$  is at most  $(p - 1)p^{m/l} / ((p - 1)p^{i-1}) = p^{n/2+1/l}$ .

Assume that  $K_0$  minimizes the stiffness among all dimensions  $i - 1$  subspaces contained in  $Y_2 \cap Q$  and assume that its stiffness equals  $l_0$ , then  $t(Y_2) \leq cp^{n/2+1}/l_0$ .

Let  $R_j, 1 \leq j \leq l_0$ , be the  $Y_1$ -spaces defining the stiffness of  $K_0$ . Then every  $Y_1$ -space  $R$  is contained in some dimension  $m - 1$  space  $R_j + K_0$ . Since  $r$  is the largest number of  $Y_1$ -spaces any dimension  $m - 1$  subspace may contain, we have  $s(Y_1) \leq r \cdot l_0$ .

Combining the results in the above two paragraphs, we have  $s(Y_1) \cdot t(Y_2) \leq r \cdot c \cdot p^{n/2+1} \leq p^{(i-1)n/2+2i-2} \cdot p^{n/2+1} = p^{i \cdot n/2+2i-1}$ . Therefore,  $E(i) \leq p^{i \cdot n/2+2i}$ .  $\square$

The  $\Theta(n^2 \log p)$  lower bound for the input partition  $\pi_0$  now follows directly from Lemmas 3.1 and 3.7.

We now prove our results for arbitrary partitions that evenly divide the input between the two agents. Since permuting the columns of a matrix does not change its rank, we only need to consider the *proper* partitions defined below.

**Definition 3.8.** An even input partition is a proper partition if and only if, for any pair of integers  $i$  and  $j$  such that  $1 \leq i < j \leq n$ , at least as many input bits from column  $i$  are assigned to the first agent as from column  $j$ .

Let  $\pi$  be any proper partition. Then  $\pi$  only differs from  $\pi_0$  on the assignment of up to  $(n^2 \log p)/2$  input bits. If the values of these input bits are fixed, then the difference between  $\pi$  and  $\pi_0$  is eliminated.

**Definition 3.9.** Given any proper partition  $\pi$ , we use  $\Delta$  to denote the set of input bit positions which are assigned differently under  $\pi_0$  and  $\pi$ . A function  $f: \Delta \rightarrow \{0, 1\}$  is called a fixing function for  $\pi$ .



No matter what fixing function we use to restrict the computation, the resulting truth matrix is a submatrix of the truth matrix defined by  $\pi_0$ . To analyze the communication complexity of this restricted computation, we select a submatrix  $T'$  in the same way we selected  $T$ . Accordingly,  $T'$  is a submatrix of  $T$ . So, the maximum size of 1-chromatic matrices in  $T'$  is still bounded according to Lemma 3.7. Therefore, it suffices to show that, for every proper partition, a fixing function exists such that the resulting matrix  $T'$  contains at least  $p^{n^2/3}$  “one” entries. This will establish Theorem 1.1.

We first prove that such fixing functions exist for some of the proper partitions.

**Definition 3.10.** A proper partition is called a *nice* partition if and only if it partitions the input bits in each column of the input matrix evenly between the agents.

In other words, if  $\pi$  is a nice partition, then  $\Delta$  contains exactly  $\frac{1}{2}n \log p$  bit positions from each column of the input matrix. We use  $\Delta_1$  to denote those bit positions of  $\Delta$  in the first  $n/2$  columns and  $\Delta_2$  to denote the other half of  $\Delta$ . Accordingly, for any fixing function  $f$  of  $\pi$ ,  $f_1$  ( $f_2$ ) is the restriction of  $f$  relative to  $\Delta_1$  ( $\Delta_2$ ).

**Definition 3.11.** Let  $X$  be any dimension  $n/2$  subspace of  $F_p^n$ , let  $\pi$  be a nice partition, and let  $f$  be a fixing function for  $\pi$ . We say  $X$  is *encodable* under  $f_1$  if and only if there is an instance of the input matrix whose bits in  $\Delta_1$  are fixed according to  $f_1$  and whose first  $n/2$  columns span  $X$ . Each of such instances is an *encoding* of  $X$  under  $f_1$ .

The notion “encodable under  $f_2$ ” is defined in a similar way. If  $X$  is encodable under  $f_1$  and  $Y$  is encodable under  $f_2$ , then we say the pair  $(X, Y)$  is encodable under  $f$ .

By Lemma 3.1, there are more than  $p^{n^2/2}$  pairs of dimensions  $n/2$  subspaces which introduce “one” entries in  $T$ . Our strategy is to show that for any nice partition there is a collection of  $p^{2\epsilon n^2}$  fixing functions such that almost all such pairs of subspaces are encodable under some function in this collection. Then, by the pigeonhole principle, there is one fixing function which yields  $p^{(1/2-2\epsilon)n^2}$  “one” entries in  $T'$ .

Assume that we are given a nice partition  $\pi$ . Our first step is to show that most dimension  $n/2$  subspaces have not too many encodings under any fixing function and hence are encodable under a large number of fixing functions.

**Definition 3.12.** Set  $\epsilon := \frac{1}{16}$ . A dimension  $n/2$  subspace is *concise* if and only if it does not have more than  $p^{2\epsilon n^2/4}$  encodings under any fixing function  $f_1$ .

Using the first  $n/2$  columns of the input matrix to encode any dimension  $n/2$  subspace  $X$ , we have  $\prod_{i=0}^{n/2-1} (p^{n/2} - p^i) \geq p^{n^2/4} \cdot e^{-2/(p-i)}$  different ways to do so. (This inequality can be proved by using some suitable inequalities appearing in the proof of Lemma 3.14.)

**Proposition 3.13.** *If  $X$  is concise, it must be encodable under more than  $p^{n^2/4-2en^{3/4}} \cdot e^{-2/(p-1)}$  fixing functions.*

**Lemma 3.14.** *Set  $\delta := p^{-e^2n^{3/2}}$ . Assume  $p \geq 8$  and  $n \geq 128$ . All the dimension  $n/2$  subspaces in  $F_p^n$  except for a  $\delta$  fraction of them, are concise.*

*Proof.* After the bit proposition in  $\Delta_1$  are fixed according to any given  $f_1$ , each column in the left half of the input matrix can only encode  $p^{n/2}$  vectors. Let  $V_i$  be the set of vectors encodable by the  $i$ th column,  $1 \leq i \leq n/2$ . To encode a dimension  $n/2$  subspace  $X$ , we must find one base vector of  $X$  in every  $V_i$ . So the number of encodings of  $X$  under  $f_1$  is not larger than  $\prod_{i=1}^{n/2} \#(X \cap V_i)$ .

If  $X$  is not concise, then there must be an  $f_1$  and an  $i$  such that  $\#(X \cap V_i) > p^{4en^{3/4}}$ . Given  $f_1$  and  $i$ , let  $U$  be the set  $\{X: \#(X \cap V_i) > p^{4en^{3/4}}\}$ . Let *Total* represent the total number of dimension  $n/2$  subspaces in  $F_p^n$ . So

$$Total = \prod_{j=0}^{n/2-1} \frac{p^n - p^j}{p^{n/2} - p^j}.$$

Each subspace in  $U$  must have some basis containing  $4en^{3/4}$  base vectors from  $V_i$ . Let us generate a basis for the subspaces in  $U$  by first choosing any  $4en^{3/4}$  linearly independent vectors in  $V_i$  and then adding  $n/2 - 4n^{3/4}$  base vectors from  $F_p^n$  one by one. The number of different basis representations is at most

$$\left( p^{n/2} \right)_{4en^{3/4}} \prod_{j=4en^{3/4}}^{n/2-1} (p^n - p^j). \tag{3.1}$$

However, each subspace in  $U$  contains at least

$$\prod_{j=0}^{4en^{3/4}-1} \frac{p^{4en^{3/4}} - p^j}{4en^{3/4} - j} \prod_{j=4en^{3/4}}^{n/2-1} (p^{n/2} - p^j) \tag{3.2}$$

sets of base vectors counted in (3.1). So,  $\#U$  is at most the quantity in (3.1) divided by the quantity in (3.2), which is

$$\prod_{j=0}^{4en^{3/4}-1} \frac{p^{n/2} - j}{p^{4en^{3/4}} - p^j} \prod_{j=4en^{3/4}}^{n/2-1} \frac{p^n - p^j}{p^{n/2} - p^j}.$$

So,  $\#U/Total$  is at most

$$\prod_{j=0}^{4en^{3/4}-1} \frac{p^{n/2} - j}{p^{4en^{3/4}} - p^j} \frac{p^{n/2} - p^j}{p^n - p^j}. \tag{3.3}$$

Note that

$$\prod_{j=0}^{m-1} (p^m - p^j) = p^{m^2} \prod_{j=0}^{m-1} (1 - p^{j-m}) = p^{m^2} \prod_{j=1}^m (1 - p^{-j}).$$

Also,

$$e^{-2p^{-j}} = 1 - 2p^{-j} + 2p^{-2j} - 4p^{-3j}/3 + \dots < 1 - p^{-j}.$$

So,

$$\begin{aligned} \prod_{j=0}^{4\epsilon n^{3/4}-1} (p^{4\epsilon n^{3/4}} - p^j)^{-1} &= p^{-16\epsilon^2 n^{3/2}} \cdot \prod_{j=1}^{4\epsilon n^{3/4}} (1 - p^{-j})^{-1} \\ &< p^{-16\epsilon^2 n^{3/2}} \cdot \prod_{j=1}^{4\epsilon n^{3/4}} e^{2p^{-j}} < p^{-16\epsilon^2 n^{3/2}} \cdot e^{2/(p-1)}. \end{aligned}$$

Since  $(p^{n/2} - p^j)/(p^n - p^j) \leq p^{-n/2}$ , we have

$$(3.3) < p^{-16\epsilon^2 n^{3/2}} e^{2/(p-1)} \cdot \prod_{j=0}^{4\epsilon n^{3/4}-1} (1 - j \cdot p^{n/2}) < p^{16\epsilon^2 n^{3/2}} e^{2/(p-1)}.$$

Because there are only  $p^{n/2}$  different ways to assign values to the bit positions in the intersection of  $\Delta_1$  and the  $i$ th column of the input matrix, for any given  $i$ , the set

$$\{S : \text{there exists } f_1 \text{ such that } \#(S \cap V_i) > p^{4\epsilon n^{3/4}}\}$$

has size less than  $(p^{n/2} p^{-16\epsilon^2 n^{3/2}} e^{2/(p-1)}) \times Total$ . Therefore the total number of nonconcise subspaces is less than a  $\delta' = (n/2)p^{n/2} p^{-16\epsilon^2 n^{3/2}} e^{2/(p-1)}$  fraction of  $Total$ . When  $p \geq 8$  and  $n \geq 128$ , we have  $\delta' < \delta$ .  $\square$

Next, we prove that almost all concise subspaces are encodable under a small collection of fixing functions.

**Lemma 3.15.** *Given any  $\Delta$ , there is a set of  $p^{\epsilon n^2}$  fixing functions,  $F_1$ , restricted to  $\Delta_1$  such that all but a  $p^{-p^n}$  fraction of concise subspaces are encodable under some  $f_1 \in F_1$ .*

*Proof.* Given a group of  $m$  concise subspaces, we count the number of pairs  $(f_1, X)$  such that  $f_1$  is a fixing function restricted to  $\Delta_1$ , and  $X$  is a concise subspace in this group encodable under  $f_1$ . By Proposition 3.13, there are at least  $m \cdot p^{n^2/4 - 2\epsilon n^{7/4}} \cdot e^{-2/(p-1)}$  such pairs. Since there are  $p^{n^2/4}$  different restricted fixing functions, one of them must be included in at least  $m \cdot p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)}$  pairs. This means we can always find a restricted fixing function  $f_1$  such that a  $p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)}$  fraction of subspaces in this group is encodable under  $f_1$ .

Starting with all the concise subspaces, we can find one restricted fixing function  $f_1$  such that a  $p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)}$  fraction of the concise subspaces are encodable under  $f_1$ . Then we can find another fixing function  $f'_1$  such that a  $p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)}$  fraction of all the remaining concise subspaces are encodable under  $f'_1$ .

We can repeat this process so that after the  $i$ th step, a  $1 - (1 - p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)})^i$  fraction of all the concise subspaces are encodable under some of the  $i$  restricted fixing functions we have found. Carrying out this process  $p^{\epsilon n^2}$  steps gives us  $F_1$ .

The fraction of concise subspaces which are not encodable under any restricted fixing function in  $F_1$  is at most  $(1 - p^{-2\epsilon n^{7/4}} \cdot e^{-2/(p-1)})p^{\epsilon n^2}$ . This quantity is smaller than  $p^{-p^n}$ .  $\square$

Lemmas 3.14 and 3.15 establish that there is a set of  $p^{\epsilon n^2}$  restricted fixing functions  $F_1$  such that almost all dimension  $n/2$  subspaces in  $F_p^n$  are encodable under some  $f_1 \in F_1$ . The exceptions are less than a  $p^{-\epsilon^2 n^{3/2}} + p^{-p^n}$  fraction of all such subspaces.

We repeat this procedure for the last  $n/2$  columns of the input matrix. This means we have  $p^{\epsilon n^2}$  restricted fixing functions for  $\Delta_2$  such that every dimension  $n/2$  subspace in  $F_p^n$  is encodable under some of these functions, with the exception of less than a  $p^{-\epsilon^2 n^{3/2}} + p^{-p^n}$  fraction of all such subspaces.

Therefore given a nice partition  $\pi'$ , we have a set of  $p^{2\epsilon n^2}$  fixing functions such that almost all pairs of dimension  $n/2$  subspaces  $(X, Y)$  are encodable under some of these functions. Since  $p^{-\epsilon^2 n^{3/2}} + p^{-p^n} \leq 2p^{-\epsilon^2 n^{3/2}}$ , the number of exceptions is less than  $4p^{-\epsilon^2 n^{3/2}} \times (Total)^2$ , where

$$Total = \prod_{j=0}^{n/2-1} \frac{p^n - p^j}{p^{n/2} - p^j}$$

is the total number of dimension  $n/2$  subspace in  $F_p^n$ . (Note that  $Total < p^{n^2/4} e^{2/(p-1)}$ , see the inequality before Proposition 3.13.)

Each of the subspace pairs corresponds to an entry in the truth matrix  $T$ . Also, there are at least  $p^{n^2/2}$  “one” entries in  $T$  by Lemma 3.1. So, there must be one fixing function which defines a submatrix,  $T'$ , of  $T$  that contains at least  $p^{(1/2-2\epsilon)n^2} - p^{(1/2-2\epsilon)n^2 - \epsilon^2 n^{3/2}} e^{O(1)}$  “one” entries. This result, together with Lemma 3.7, establishes the following lemma by an application of Yao’s lower-bound method.

**Lemma 3.16.** *Under any nice partition  $\pi'$ , the singularity test over the finite field  $F_p$  requires at least  $(n^2 \log p)/8 - n \log p - O(1)$  bits of communication.*

Now we need to show that  $\Theta(n^2 \log p)$  bits of communication are required for any proper partition  $\pi$ .

**Definition 3.17.** (Continuing from Definition 3.9). For  $1 \leq i \leq n$ , let  $\delta_i$  be the set of input bit positions in  $\Delta$  which belong to the  $i$ th column of the input matrix.

In case  $\#\delta_i \leq (n \log p)/2$  for all  $i$ , we can proceed as in the nice partition case by picking  $(n \log p)/2$  bit positions to be fixed in every column, which include all positions in  $\delta_i$ ,  $1 \leq i \leq n$ . The  $\Theta(n^2 \log p)$  bound follows directly.

Now assume that  $\#\delta_i > (n \log p)/2$  for some  $i$ . Then either  $\#\delta_{n/2} > (n \log p)/2$  or  $\#\delta_{n/2+1} > (n \log p)/2$ , exclusively (Definition 3.8). We need to address only one of these cases, say  $\#\delta_{n/2} > (n \log p)/2$ .

Let  $m \leq n/2$  be the index such that  $\#\delta_m > (n \log p)/2$  and  $\#\delta_{m-1} \leq (n \log p)/2$ . For each  $m \leq i \leq n/2$ , define  $d_i$  to be  $(\#\delta_i) - (n \log p)/2$ . Pick arbitrary  $d_i$  bit positions from each  $\delta_i$ ,  $m \leq i \leq n/2$ , and call them characteristic positions of  $\pi$ . Let  $D = \sum_{i=m}^{n/2} d_i$  denote the total number of these characteristic positions.

The proper partition  $\pi$  can be reduced to a nice partition  $\pi'$  by reassigning all the characteristic positions to the first agent and compensate the second agent with positions in column 1 through column  $(m - 1)$  which are originally assigned to the first agent and which are to be fixed. Therefore, the amount of communication required under  $\pi$  can be less than that required under  $\pi'$  by at most  $D$  bits.

**Lemma 3.18.** *The number of characteristic positions,  $D$  of any proper partition is less than  $(\frac{1}{16})n^2 \log p$ .*

*Proof.* By Definition 3.8,  $d_i \leq d_{i+1}$  for  $m \leq i < n/2$ . Let  $l = n/2 - m + 1$ . Then  $D \leq l \cdot d_{n/2}$ . The second agent receives at least  $\#\delta_{n/2}$  positions from each column in the right half of the input matrix. The total number of bit positions assigned to the second agent is at least  $\sum_{i=m}^{n/2} \delta_i + (n/2)(\#\delta_{n/2})$ . This value cannot exceed  $(n^2 \log p)/2$ . So we have the following inequality:

$$\frac{nl \log p}{2} + D + \frac{n^2 \log p}{4} + \frac{n}{2}d_{n/2} \leq \frac{n^2 \log p}{2}. \tag{3.4}$$

Solve for  $d_{n/2}$  in (3.4) and we get

$$d_{n/2} \leq \frac{n \log p}{2} - l \log p - \frac{2D}{n}. \tag{3.5}$$

Since  $D \leq l \cdot d_{n/2}$ , multiplying both sides of the inequality (4.5) with  $l$  gives us

$$D \leq \frac{l(n - 2l)}{(2l + n)2n} n^2 \log p. \tag{3.6}$$

Assume  $l = cn$ . Solve (3.6) for the maximum of  $D$ . We get  $D = (\frac{3}{4} - \sqrt{2}/2)n^2 \log p$ , when  $c = (-1 + \sqrt{2})/2$ . This maximum value is less than  $(\frac{1}{16})n^2 \log p$ . □

Therefore, by Lemmas 3.16 and 3.18, the communication complexity of the singularity test over the finite field  $F_p$  is at least  $(n^2 \log p)/16 - n \log p - O(1)$ , which is in  $\Theta(n^2 \log p)$ .

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