# On the upper bounds of the minimum number of rows of disjunct matrices 

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#### Abstract

A 0-1 matrix is $d$-disjunct if no column is covered by the union of any $d$ other columns. A 0-1 matrix is $(d ; z)$-disjunct if for any column $C$ and any $d$ other columns, there exist at least $z$ rows such that each of them has value 1 at column $C$ and value 0 at all the other $d$ columns. Let $t(d, n)$ and $t(d, n ; z)$ denote the minimum number of rows required by a $d$-disjunct matrix and a $(d ; z)$-disjunct matrix with $n$ columns, respectively. We give a very short proof for the currently best upper bound on $t(d, n)$. We also generalize our method to obtain a new upper bound on $t(d, n ; z)$.


Keywords Disjunct matrices • Cover free families • Superimposed codes

## 1 Introduction

A 0-1 matrix is $d$-disjunct if no column is covered by the union of any $d$ other columns, by union we mean the bitwise boolean sum of these $d$ column vectors. In other words, a 0-1 matrix is called $d$-disjunct if for any column $C$ and any $d$ other columns, there

[^0]exists at least one row such that the row has value 1 at column $C$ and value 0 at all $d$ other columns. The same structure is also called cover free family $[9,10,15]$ in combinatorics, and superimposed code $[6,8,12]$ in information theory. It is called a $d$-disjunct matrix in group testing $[4,11,13]$. A $0-1$ matrix is $(d ; z)$-disjunct $[8,13]$ if for any column $C$ and any $d$ other columns, there exist at least $z$ rows such that each of them has value 1 at column $C$ and value 0 at all the other $d$ columns. Thus, $d$-disjunct is $(d ; 1)$-disjunct. Besides other applications, $d$-disjunct and $(d ; z)$-disjunct matrices form the basis for error-free and error-tolerant nonadaptive group testing algorithms, respectively. These algorithms have applications in many practical areas such as DNA library screening $[2-4,14]$ and multi-access communications [16], etc.

Let $t(d, n)$ denote the minimum number of rows required by a $d$-disjunct matrix with $n$ columns. The bounds on $t(d, n)$ have been extensively studied in the fields of combinatorics, information theory, and group testing, using different terminologies. For lower bounds, $t(d, n)=\Omega\left(\frac{d^{2} \log n}{\log d}\right)[7,10,15]$ (throughout the paper $\log$ is of base 2 if no base is specified). In particular, D'yachkov and Rykov [7] proved that $t(d, n) \geq \frac{d^{2}}{2 \log d}(1+o(1)) \log n$, which is the best lower bound so far. For upper bounds on $t(d, n)$, it is known that $t(d, n)=O\left(d^{2} \log n\right)$ [8,11]. In [11] (also see [4, p. 57]), Hwang and Sós gave a greedy type construction which results in $t \times n$ $d$-disjunct matrices with $t \leq 16 d^{2}\left(1-\log _{3} 2+\left(\log _{3} 2\right) \log _{2} n\right)$. In [8], D'yachkov et al. obtained the following asymptotic upper bound on $t(d, n)$ with a rather involved proof, which is currently the best.

Theorem 1.1 (D'yachkov et al. [8]) For $d$ constant and $n \rightarrow \infty, t(d, n) \leq \frac{d}{A_{d}}[1+$ $o(1)] \log n$, where $A_{d}=\max _{0 \leq p \leq 1} \max _{0 \leq P \leq 1}\left\{-(1-P) \log \left(1-p^{d}\right)+d\left[P \log \frac{p}{P}+\right.\right.$ $\left.\left.(1-P) \log \frac{1-p}{1-P}\right]\right\}$. Moreover, $A_{d} \rightarrow \frac{1}{d \log e}$ as $d \rightarrow \infty$.

For $(d ; z)$-disjunct matrices, let $t(d, n ; z)$ denote the minimum number of rows required by a $(d ; z)$-disjunct matrix with $n$ columns. For given $d$ and $z$, D' yachkov et al. [8] studied $\lim _{n \rightarrow \infty} \frac{\log n}{t}$ among others, and they proved that $t(d, n ; z) \geq c\left[\frac{d^{2} \log n}{\log d}+\right.$ $(z-1) d]$ where $c$ is a constant.

In this paper, by using the concept of $q$-ary ( $d, 1$ )-disjunct matrices $[4,5]$ and the probabilistic method (see, e.g., [1]), we give a very short proof for the currently best upper bound on $t(d, n)$. In contrast to the previous result in [8] (Theorem 1.1) which is an asymptotic upper bound, our upper bound on $t(d, n)$ does not contain the asymptotic term $o(1)$. Also, we generalize our method to obtain a new upper bound on $t(d, n ; z)$. Since our new proof is very short and concise, we hope that it can shed new light on this old problem and stimulate new research on it.

## 2 Upper bounds on $t(d, n)$

In this section we prove the following theorem.
Theorem 2.1 For $n>d \geq 1, t(d, n) \leq \frac{d+1}{B_{d}} \log n$, where $B_{d}=\max _{q>1} \frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$. Moreover, $B_{d} \rightarrow \frac{1}{d \log e}$ as $d \rightarrow \infty$.

Before the proof, we first introduce the concept of $q$-ary $(d, 1)$-disjunct matrix. A matrix is called $q$-ary $(d, 1)$-disjunct if it is $q$-ary, and for any column $C$ and any set $D$ of $d$ other columns, there exists an element in $C$ such that the element does not appear in any column of $D$ in the same row.

As described in $[4,5]$, one can transform a $q$-ary $(d, 1)$-disjunct matrix $M$ to a (binary) $d$-disjunct matrix $M^{\prime}$ as follows. Replace each row $R_{i}$ of $M$ by several rows indexed with entries of $R_{i}$. For each entry $x$ of $R_{i}$, the row with index $x$ is obtained from $R_{i}$ by turning all $x$ 's into 1 's and all others into 0 's. From this transformation, we have the following theorem which is useful in our proof.
Theorem 2.2 (Theorem 3.6.1 in [4]) A $t \times n q$-ary ( $d, 1$ )-disjunct matrix $M$ yields a $t^{\prime} \times n d$-disjunct matrix $M^{\prime}$ with $t^{\prime} \leq t q$.
Now we are ready to prove Theorem 2.1.

Proof of Theorem 2.1 Given $n>d \geq 1$, first construct a random $t \times n q$-ary $(q>1)$ matrix $M$ with each entry assigned randomly and uniformly from $\{1,2, \ldots, q\}$, where $q$ and $t$ will be specified later. For each column $C$ and a set $D$ of $d$ other columns, for each element $c_{i}(i=1,2, \ldots, t)$ of $C$, the probability that $c_{i}$ appears in some column of $D$ in the same row is $1-\left(1-\frac{1}{q}\right)^{d}$. Thus the probability that every element of $C$ appears in some column of $D$ in the same row is $\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t} . M$ is not $(d, 1)$-disjunct if and only if there exist a column $C$ and a set $D$ of $d$ other columns such that the above holds. Therefore, the probability that $M$ is not $(d, 1)$-disjunct is no more than $(n-d)\binom{n}{d}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t}$.

We try to minimize $t q$, the number of rows of the $d$-disjunct matrix $M^{\prime}$ as in Theorem 2.2, under the condition that $q$ and $t$ satisfy

$$
\begin{equation*}
n^{d+1}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t} \leq 1 \tag{2.1}
\end{equation*}
$$

Notice that Eq. (2.1) implies $(n-d)\binom{n}{d}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t}<1$, thus the probability that $M$ is $(d, 1)$-disjunct is greater than zero. Therefore, by probabilistic argument Eq. (2.1) implies the existence of a $t \times n q$-ary $(d, 1)$-disjunct matrix, and so a $d$-disjunct matrix with $n$ columns and at most $t q$ rows.

To satisfy Eq. (2.1), let $t=\frac{(d+1) \log n}{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}$. Define $B_{d}(q)=\frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$, then $t q=\frac{(d+1) \log n}{B_{d}(q)}$. Let $q_{0}$ be the point that maximizes $B_{d}(q)$, and let $B_{d}=B_{d}\left(q_{0}\right)$ (one can estimate that $q_{0}=\Theta(d)$ and $B_{d}=\Theta\left(\frac{1}{d}\right)$, since the proof here can stand alone without this observation, we put it in appendix). By assigning $q=q_{0}$, we obtain

$$
t(d, n) \leq\left.(t q)\right|_{q=q_{0}}=\frac{(d+1) \log n}{B_{d}\left(q_{0}\right)}=\frac{(d+1) \log n}{B_{d}} .
$$

Finally, we estimate $B_{d}$ as $d \rightarrow \infty$. Since $\left(1-\frac{1}{q}\right)^{q}<\frac{1}{e}$ for $q>1,\left(1-\frac{1}{q}\right)^{d}$ $<\left(\frac{1}{e}\right)^{\frac{d}{q}}=e^{-\frac{d}{q}}$, and $-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]<-\log \left(1-e^{-\frac{d}{q}}\right)$. It follows that
$B_{d}(q)=\frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}<\frac{-\log \left(1-e^{-\frac{d}{q}}\right)}{q}=\frac{1}{d \ln 2}\left[-\frac{d}{q} \ln \left(1-e^{-\frac{d}{q}}\right)\right]$. Let $x=e^{-\frac{d}{q}}$, then $-\frac{d}{q}=\ln x$, and $B_{d}(q)<\frac{1}{d \ln 2} \ln x \ln (1-x)$. Since $\ln x \ln (1-x)$ achieves its maximum at $x=\frac{1}{2}$, we obtain $B_{d}(q)<\frac{\ln 2}{d}$ for $q>1$. Thus $B_{d}<\frac{\ln 2}{d}$ for $d \geq 1$. On the other hand, when $q$ satisfies $\left(1-\frac{1}{q}\right)^{d}=\frac{1}{2}$, as $d \rightarrow \infty$, it is easy to see that $\frac{q}{d} \rightarrow \frac{1}{\ln 2}$, and $B_{d}(q)=\frac{1}{q} \rightarrow \frac{\ln 2}{d}$. Therefore, as $d \rightarrow \infty, B_{d} \rightarrow \frac{\ln 2}{d}=\frac{1}{d \log e}$.

## 3 New upper bounds on $t(d, n ; z)$

In this section, we generalize the above method to obtain new upper bounds for $(d ; z)$ disjunct matrices. We establish the following theorem.
Theorem 3.1 For $d, z$ constants, and $n \rightarrow \infty, t(d, n ; z) \leq \frac{d+1}{B_{d}} \log n+\frac{z}{B_{d}} \log$ $\log n+O(1)$, where $B_{d}=\max _{q>1} \frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$. Moreover, $B_{d} \rightarrow \frac{1}{d \log e}$ as $d \rightarrow \infty$.

A $q$-ary matrix is called $(d, 1 ; z)$-disjunct if for any column $C$ and any set $D$ of $d$ other columns, there exists at least $z$ elements in $C$ such that each of these elements does not appear in any column of $D$ in the same row. Clearly, by using the same method mentioned above, one can transform a $t \times n q$-ary $(d, 1 ; z)$-disjunct matrix to a $(d ; z)$-disjunct matrix with $n$ columns and at most $t q$ rows.

Proof of Theorem 3.1 For given $n, d$ and $z$, similarly we construct a random $t \times n q$-ary $(q>1)$ matrix $M$ with each entry assigned randomly and uniformly from $\{1,2, \ldots, q\}$, $q$ and $t$ will be specified later. For each column $C$ and a set $D$ of $d$ other columns, for each element $c_{i}$ of $C$, the probability that $c_{i}$ appears in some column of $D$ in the same row is $1-\left(1-\frac{1}{q}\right)^{d}$. Thus the probability that there exist $t-z+1$ elements of $C$ such that each of them appears in some column of $D$ in the same row is at most $\binom{t}{t-z+1}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t-z+1}=\binom{t}{z-1}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t-z+1} . M$ is not $(d, 1 ; z)$-disjunct if and only if there exists a column $C$ and a set $D$ of $d$ other columns such that the above holds. Therefore, the probability that $M$ is not $(d, 1 ; z)$-disjunct is no more than $(n-d)\binom{n}{d}\binom{t}{z-1}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t-z+1}$.

We want to minimize $t q$, the number of rows of the corresponding $(d ; z)$-disjunct matrix, under the condition that

$$
\begin{equation*}
n^{d+1} t^{z}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t-z} \leq 1 \tag{3.1}
\end{equation*}
$$

Notice that Eq. (3.1) implies $(n-d)\binom{n}{d}\binom{t}{z-1}\left[1-\left(1-\frac{1}{q}\right)^{d}\right]^{t-z+1}<1$. Thus the probability that $M$ is $(d, 1 ; z)$-disjunct is greater than zero, which similarly implies the existence of a $t \times n q$-ary $(d, 1 ; z)$-disjunct matrix, and a $(d ; z)$-disjunct matrix with $n$ columns and at most $t q$ rows.

Let $q_{0}$ be the point that maximizes $B_{d}(q)=\frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$. Assign $q=q_{0}$. To satisfy Eq. (3.1), which is equivalent to $(d+1) \log n+z \log t \leq-(t-z) \log [1-$ $\left.\left(1-\frac{1}{q_{0}}\right)^{d}\right]$, let $t=\frac{(d+1) \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+z+t_{1}$. Then, $t_{1}$ should satisfy

$$
\begin{equation*}
z \log \left\{\frac{(d+1) \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+z+t_{1}\right\} \leq-t_{1} \log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right] \tag{3.2}
\end{equation*}
$$

Let $t_{1}=\frac{z \log \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+t_{2}$, from Eq. (3.2), $t_{2}$ should satisfy that

$$
\begin{gather*}
\frac{z}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]} \log \left\{\frac{(d+1)}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}\right. \\
\left.+\frac{1}{\log n}\left(\frac{z \log \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+z+t_{2}\right)\right\} \leq t_{2} . \tag{3.3}
\end{gather*}
$$

For $d$ and $z$ constants (thus $q_{0}$ is also constant), as $n \rightarrow \infty$, the minimum value of $t_{2}$ satisfying Eq. (3.3) is $t_{2}=\frac{z}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]} \log \frac{(d+1)}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}=O(1)$. Thus,

$$
t=\frac{(d+1) \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+\frac{z \log \log n}{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}+O(1)
$$

satisfies Eq. (3.1) (where the constant term $z$ in $t$ is absorbed in $\mathrm{O}(1)$ ). Therefore, the number of rows of the corresponding $(d ; z)$-disjunct matrix is at most

$$
t q_{0}=\frac{d+1}{B_{d}} \log n+\frac{z}{B_{d}} \log \log n+O(1)
$$

where $B_{d}=B_{d}\left(q_{0}\right)=\max _{q>1} \frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$. Also, $B_{d} \rightarrow \frac{1}{d \log e}$ as $d \rightarrow \infty$, as proved in Theorem 2.1.

## Appendix A: Estimating $\boldsymbol{q}_{0}$ and $\boldsymbol{B}_{\boldsymbol{d}}$

Lemma A. 1 Given $d \geq 1$, let $q_{0}=q_{0}(d)$ be the point that maximizes $B_{d}(q)=$ $\frac{-\log \left[1-\left(1-\frac{1}{q}\right)^{d}\right]}{q}$ for $q>1$. Then, as $d \rightarrow \infty, q_{0}(d)=\Theta(d)$, and $B_{d}=B_{d}\left(q_{0}\right)=$ $\Theta\left(\frac{1}{d}\right)$.

Proof Notice that if $q_{1}$ satisfies $\left(1-\frac{1}{q_{1}}\right)^{d}=\frac{1}{2}$, then $q_{1}=\Theta(d)$ since $\frac{q_{1}}{d} \rightarrow \frac{1}{\ln 2}$ as $d \rightarrow \infty$. Moreover, $B_{d}\left(q_{1}\right)=\frac{1}{q_{1}}=\Theta\left(\frac{1}{d}\right)$. We prove the lemma by contradiction. First assume that $q_{0}=O(d)$ does not hold, that is, for any $c>0$ and any $d_{0}>0$,
there exists $d>d_{0}$ such that $q_{0}(d)>c d$. Then, since $\frac{q_{0}}{d}>c$, as $c \rightarrow \infty, B_{d}\left(q_{0}\right) d=$ $\frac{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}{q_{0}} d \sim \frac{-\log \left[1-\left(1-\frac{d}{q_{0}}\right)\right]}{q_{0}} d=\frac{\log \frac{q_{0}}{d}}{\frac{q_{d}}{d}}=o(1)$ (here by $a \sim b$ we mean that $\left.\lim _{c \rightarrow \infty} \frac{a}{b}=1\right)$. Thus, $B_{d}\left(q_{0}\right)=\frac{o(1)}{d}$. It contradicts since $q_{0}$ is the maximum point of $B_{d}(q)$ and $B_{d}\left(q_{1}\right)=\Theta\left(\frac{1}{d}\right)$ with $\left(1-\frac{1}{q_{1}}\right)^{d}=\frac{1}{2}$. On the other hand, assume that $q_{0}=\Omega(d)$ does not hold, that is, for any $c>0$ and any $d_{0}>0$, there exists $d>d_{0}$ such that $q_{0}(d)<c d$. Then, $B_{d}\left(q_{0}\right) d=\frac{-\log \left[1-\left(1-\frac{1}{q_{0}}\right)^{d}\right]}{q_{0}} d=\frac{-\ln \left\{1-\left[\left(1-\frac{1}{q_{0}}\right)^{q_{0}}\right]^{\frac{d}{q_{0}}}\right\}}{q_{0} \ln 2} d$. Since $0<\left(1-\frac{1}{q_{0}}\right)^{q_{0}}<\frac{1}{e}$ for $q_{0}>1$, as $c \rightarrow 0, \frac{d}{q_{0}}>\frac{1}{c} \rightarrow \infty$, and $\left[\left(1-\frac{1}{q_{0}}\right)^{q_{0}}\right]^{\frac{d}{q_{0}}}<$ $e^{-\frac{d}{q_{0}}} \rightarrow 0$. Thus $B_{d}\left(q_{0}\right) d \sim \frac{\left[\left(1-\frac{1}{q_{0}}\right)^{q_{0}}\right]^{\frac{d}{q_{0}}}}{q_{0} \ln 2} d=\frac{1}{\ln 2} \frac{d}{q_{0}}\left[\left(1-\frac{1}{q_{0}}\right)^{q_{0}}\right]^{\frac{d}{q_{0}}}<\frac{1}{\ln 2} \frac{d}{q_{0}} e^{-\frac{d}{q_{0}}}=$ $o(1)$, which also contradicts (here by $a \sim b$ we mean that $\lim _{c \rightarrow 0} \frac{a}{b}=1$ ). Therefore, $q_{0}(d)=\Theta(d)$. Then, $\left(1-\frac{1}{q_{0}}\right)^{d}<1$ is $\Theta(1)$, and thus $B_{d}\left(q_{0}\right)=\frac{\Theta(1)}{q_{0}}=\Theta\left(\frac{1}{d}\right)$.

## References

1. Alon, N., Spencer, J.H.: The Probabilistic Method. Wiley, New York (1992)
2. Balding, D.J., Bruno, W.J., Knill, E., Torney, D.C.: A comparative survey of non-adaptive pooling designs. In: Genetic Mapping and DNA Sequencing, pp. 133-154. Springer, New York (1996)
3. Bruno, W.J., Balding, D.J., Knill, E., Bruce, D.C., Doggett, N.A., Sawhill, W.W., Stallings, R.L., Whittaker, C.C., Torney, D.C.: Efficient pooling designs for library screening. Genomics 26, 21-30 (1995)
4. Du, D.Z., Hwang, F.K.: Pooling Designs and Nonadaptive Group Testing: Important Tools for DNA Sequencing. World Scientific, Singapore (2006)
5. Du, D.Z., Hwang, F.K., Wu, W., Znati, T.: New construction for transversal design. J. Comput. Biol. 13, 990-995 (2006)
6. D'yachkov, A.G., Macula, A.J., Rykov, V.V.: New constructions of superimposed codes. IEEE Trans. Inform. Theory 46, 284-290 (2000)
7. D'yachkov, A.G., Rykov, V.V.: Bounds of the length of disjunct codes. Problems Control Inform. Theory 11, 7-13 (1982)
8. D'yachkov, A.G., Rykov, V.V., Rashad, A.M.: Superimposed distance codes. Problems Control Inform. Theory 18, 237-250 (1989)
9. Erdös, P., Frankl, P., Füredi, Z.: Families of finite sets in which no set is covered by the union of r others. Israel J. Math. 51, 79-89 (1985)
10. Füredi, Z.: On $r$-cover-free families. J. Comb. Theory Ser. A 73, 172-173 (1996)
11. Hwang, F.K., Sós, V.T.: Non-adaptive hypergeometric group testing. Studia Scient. Math. Hungarica. 22, 257-263 (1987)
12. Kautz, W.H., Singleton, R.C.: Nonrandom binary superimposed codes. IEEE Trans. Inform. Theory 10, 363-377 (1964)
13. Macula, A.J.: Error-correcting nonadaptive group testing with $d^{e}$-disjunct matrices. Discrete Appl. Math. 80, 217-222 (1997)
14. Ngo, H.Q., Du, D.Z.: A survey on combinatorial group testing algorithms with applications to DNA library screening. DIMACS: Series in Discrete Mathematics and Theoretical Computer Science, vol. 55, pp. 171-182. American Mathematical Society, Providence (2000)
15. Ruszinkó, M.: On the upper bound of the size of the $r$-cover-free families. J. Comb. Theory Ser. A 66, 302-310 (1994)
16. Wolf, J.K.: Born again group testing: multiaccess communications. IEEE Trans. Inform. Theory IT-31, 185-191 (1998)

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