# Improved Semidefinite Programming Bound on Sizes of Codes

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#### Abstract

Let A(n,d) (respectively A(n,d,w)) be the maximum possible number of codewords in a binary code (respectively binary constant-weight w code) of length n and minimum Hamming distance at least d. By adding new linear constraints to Schrijver's semidefinite programming bound, which is obtained from block-diagonalising the Terwilliger algebra of the Hamming cube, we obtain two new upper bounds on A(n,d), namely  $A(18,8) \le 71$  and  $A(19,8) \le 131$ . Twenty three new upper bounds on A(n,d,w) for  $n \le 28$  are also obtained by a similar way.

#### **Index Terms**

Binary codes, binary constant-weight codes, linear programming, semidefinite programming, upper bound.

### I. INTRODUCTION

Let  $\mathcal{F} = \{0, 1\}$  and let *n* be a positive integer. The *(Hamming) distance* between two vectors in  $\mathcal{F}^n$  is the number of coordinates where they differ. The *(Hamming) weight* of a vector in  $\mathcal{F}^n$  is the distance between it and the zero vector. The *minimum distance* of a subset of  $\mathcal{F}^n$  is the smallest distance between any two different vectors in that subset. An (n, d) code is a subset of  $\mathcal{F}^n$  having minimum distance  $\geq d$ . If  $\mathcal{C}$  is an (n, d) code, then an element of  $\mathcal{C}$  is called a *codeword* and the number of codewords in  $\mathcal{C}$  is called the *size* of  $\mathcal{C}$ .

The largest possible size of an (n, d) code is denoted by A(n, d). The problem of determining the exact values of A(n, d) is one of the most fundamental problems in combinatorial coding theory. Among upper bounds on A(n, d), Delsarte's linear programming bound is quite powerful (see [1] and [2]). This bound is obtained from block-diagonalising the Bose-Mesner algebra of  $\mathcal{F}^n$ . In 2005, by block-diagonalising the Terwilliger algebra (which contains the Bose-Mesner algebra) of  $\mathcal{F}^n$ , Schrijver gave a semidefinite programming bound [3]. This bound was shown to be stronger than or as good as Delsarte's linear programming bound. In fact, eleven new upper bounds on A(n, d) were obtained in the paper for  $n \leq 28$ . In 2002, Mounits, Etzion, and Litsyn added more linear constraints to Delsarte's linear programming bound and obtained new upper bounds on A(n, d) [4]. In this paper, we construct new linear constraints and show that these linear constraints improve Schrijver's semidefinite programming bound.

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Among improved upper bounds on A(n,d) for  $n \le 28$ , there are two new upper bounds, namely  $A(18,8) \le 71$ and  $A(19,8) \le 131$ .

An (n, d, w) constant-weight code is an (n, d) code such that every codeword has weight w. Let A(n, d, w) be the largest possible size of an (n, d, w) constant-weight code. The problem of determining the exact values of A(n, d, w) has its own interest. Upper bounds on A(n, d, w) can even help to improve upper bounds on A(n, d) (for example, see [4], [2]). There are also Delsarte's linear programming bound and Schrijver's semidefinite programming bound on A(n, d, w) [1], [3]. In 2000, Agrell, Vardy, and Zeger added new linear constraints to Delsarte's linear programming bound and improved several upper bounds on A(n, d, w) [5]. More linear constraints that improve upper bounds on A(n, d, w) can be found in [6]. In this paper, we add further new linear constraints to Schrijver's semidefinite programming bound on A(n, d, w) for  $n \le 28$ .

## II. UPPER BOUNDS ON A(n, d)

In this section, we improve upper bounds on A(n, d) by adding more linear constraints to Schrijver's semidefinite programming bound, which is obtained from block-diagonalising the Terwilliger algebra of the Hamming cube  $\mathcal{F}^n$ . For more details about Schrijver's semidefinite programming bound, see [3].

## A. General Definition of A(n, d) and A(n, d, w)

We first give a general definition. Let n and d be positive integers. For a finite (possibly empty) set  $\Lambda = \{(X_i, d_i)\}_{i \in I}$ , where each  $X_i$  is a vector in  $\mathcal{F}^n$  and each  $d_i$  is a nonnegative integer, we define

$$A(n, \Lambda, d) = \text{maximum possible number of}$$
  

$$codewords \text{ in a binary code of}$$
  

$$length n \text{ and minimum distance}$$
  

$$\geq d \text{ such that each codeword is}$$
  
at distance  $d_i$  from  $X_i, \forall i \in I.$  (1)

1)  $|\Lambda| = 0$ : If  $\Lambda$  is empty, then we get the usual definition of A(n, d).

2)  $|\Lambda| = 1$ : If  $\Lambda$  contains only one element, says  $(X_1, d_1)$ , then  $A(n, \Lambda, d)$  is the maximum possible number of codewords in a binary code of length n and minimum distance  $\geq d$  such that each codeword is at distance  $d_1$  from  $X_1$ . By translation, we may assume that  $X_1$  is the zero vector so that each codeword has weight  $d_1$ . Therefore,

$$A(n,\Lambda,d) = A(n,d,w),$$
(2)

where  $w = d_1$ .

A  $(w_1, n_1, w_2, n_2, d)$  doubly-constant-weight code is an  $(n_1 + n_2, d, w_1 + w_2)$  constant-weight code such that every codeword has exactly  $w_1$  ones on the first  $n_1$  coordinates (and hence has exactly  $w_2$  ones on the last  $n_2$ coordinates). Let  $T(w_1, n_1, w_2, n_2, d)$  be the largest possible size of a  $(w_1, n_1, w_2, n_2, d)$  doubly-constant-weight code. Agrell, Vardy, and Zeger showed in [5] that upper bounds on  $T(w_1, n_1, w_2, n_2, d)$  can help improving upper bounds on A(n, d, w). In our result, upper bounds on  $T(w_1, n_1, w_2, n_2, d)$  will be used to improve upper bounds on A(n, d). As A(n, d) and A(n, d, w),  $T(w_1, n_1, w_2, n_2, d)$  is also a special case of  $A(n, \Lambda, d)$ .

3)  $|\Lambda| = 2$ : If  $\Lambda$  contains two elements, then the following proposition shows that  $A(n, \Lambda, d)$  is exactly  $T(w_1, n_1, w_2, n_2, d)$ .

Proposition 1: If  $\Lambda = \{(X_1, d_1), (X_2, d_2)\}$ , then

$$A(n,\Lambda,d) = T(w_1, n_1, w_2, n_2, d),$$
(3)

where  $n_1 = d(X_1, X_2), n_2 = n - n_1, w_1 = \frac{1}{2}(d_1 - d_2 + n_1)$ , and  $w_2 = \frac{1}{2}(d_1 + d_2 - n_1)$ .

*Proof:* Let  $n_1 = d(X_1, X_2)$  and  $n_2 = n - n_1$ . By translation, we may assume that  $X_1$  is the zero vector. Hence,  $d(X_1, X_2) = wt(X_2)$ . Let Y be a vector at distance  $d_1$  from  $X_1$  and at distance  $d_2$  from  $X_2$ . By rearranging the coordinates, we may assume that

		$n_1$	$n_2$
$X_1$	=	$\overbrace{0\cdots0}0\cdots0$	$\overbrace{0\cdots0}0\cdots0$
$X_2$	=	$1 \cdots 11 \cdots 1$	$0 \cdots 00 \cdots 0$
Y	=	$0\cdots 0\underbrace{1\cdots 1}$	$\underbrace{1\cdots 1}_{0\cdots 0}$
		$w_1$	$w_2$

Since  $X_1$  is the zero vector, we have

$$w_1 + w_2 = wt(Y) = d(Y, X_1) = d_1.$$
(4)

Also,

$$(n_1 - w_1) + w_2 = d(Y, X_2) = d_2.$$
(5)

(4) and (5) give  $w_1 = \frac{1}{2}(d_1 - d_2 + n_1)$  and  $w_2 = \frac{1}{2}(d_1 + d_2 - n_1)$ .

4)  $|\Lambda| \ge 3$ : It becomes more complicated when  $\Lambda$  contains more than two elements. We consider a very special case when  $|\Lambda| = 4$ , which will be used in our improving upper bounds on A(n, d, w) in Section III. Suppose that  $\Lambda = \{(X_1, d_1), (X_2, d_2), (X_3, d_3), (X_4, d_4)\}$  satisfies the following conditions.

- $X_1$  is the zero vector (which can always be assumed).
- $X_2$  and  $X_3$  have the same weight  $d_1$ .
- $X_4 = X_2 + X_3$ .

Then  $A(n, \Lambda, d) = T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$ , where  $w_i$  and  $n_i$   $(1 \le i \le 4)$  are determined in the next proposition. The definition of  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$  is similar to that of  $T(n_1, w_1, n_2, w_2, d)$  (it is the largest possible size of a  $(\sum_{i=1}^4 n_i, d)$  code such that on each codeword there are exactly  $w_i$  ones on the  $n_i$  coordinates  $(1 \le i \le 4)$ ).

Proposition 2: Suppose that  $\Lambda = \{(X_i, d_i)\}_{i=1}^4$  satisfies  $X_1$  is the zero vector,  $wt(X_2) = wt(X_3) = d_1$ , and  $X_4 = X_2 + X_3$ . Then

$$A(n,\Lambda,d) = T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d),$$
(6)

where  $n_1 = n_3 = \frac{1}{2}d(X_2, X_3), n_2 = d_1 - n_1, n_4 = n - n_1 - n_2 - n_3$ 

$$w_{1} = \frac{1}{4}(d_{1} - d_{2} + d_{3} - d_{4}) + \frac{1}{2}n_{1},$$

$$w_{2} = \frac{1}{4}(d_{1} - d_{2} - d_{3} + d_{4}) + \frac{1}{2}n_{2},$$

$$w_{3} = \frac{1}{4}(d_{1} + d_{2} - d_{3} - d_{4}) + \frac{1}{2}n_{3},$$

$$w_{4} = \frac{1}{4}(d_{1} + d_{2} + d_{3} + d_{4}) + \frac{1}{2}(n_{4} - n).$$

*Proof:* Suppose that Z is a vector at distance  $d_i$  from  $X_i$   $(1 \le i \le 4)$ . By rearranging the coordinates, we may assume the following.

$$X_{2} = \underbrace{1 \cdots \cdots 1}_{w_{1}} \underbrace{1 \cdots \cdots 1}_{w_{2}} \underbrace{1 \cdots \cdots 1}_{w_{1}} \underbrace{1 \cdots \cdots 1}_{w_{2}} \underbrace{1 \cdots \cdots 1}_{w_{2}} \underbrace{1 \cdots 1}_{w_{3}} \underbrace{1 \cdots 1}_{w_{4}} \underbrace{1$$

Let  $n_1, n_2, n_3, n_4$  be as in the above figure. Since  $n_1 + n_3 = d(X_2, X_3)$  and  $X_2, X_3$  have the same weight,  $n_1 = n_3 = \frac{1}{2}d(X_2, X_3)$ . Now  $n_1 + n_2 = wt(X_2) = d_1$ . Therefore,  $n_2 = d_1 - n_1$  and  $n_4 = n - n_1 - n_2 - n_3$ . We have

$$w_1 + w_2 + w_3 + w_4 = wt(Z) = d(Z, X_1) = d_1$$
  

$$(n_1 - w_1) + (n_2 - w_2) + w_3 + w_4 = d(Z, X_2) = d_2$$
  

$$w_1 + (n_2 - w_2) + (n_3 - w_3) + w_4 = d(Z, X_3) = d_3$$
  

$$(n_1 - w_1) + w_2 + (n_3 - w_3) + w_4 = d(Z, X_4) = d_4$$

Solving these equations, we get  $w_i$   $(1 \le i \le 4)$  as desired.

## B. Schrijver's Semidefinite Programming Bound on A(n, d)

Let  $\mathcal{P}$  be the collection of all subsets of  $\{1, 2, ..., n\}$ . Each vector in  $\mathcal{F}^n$  can be identified with its support (the support of a vector is the set of coordinates at which the vector has nonzero entries). With this identification, a code is a subset of  $\mathcal{P}$  and the (Hamming) distance between two subsets X and Y in  $\mathcal{P}$  is  $d(X, Y) = |X\Delta Y|$ . Let  $\mathcal{C}$  be an (n, d) code. For each i, j, and t, define

$$x_{i,j}^{t} = \frac{1}{|\mathcal{C}| \binom{n}{(i-t,j-t,t)}} \lambda_{i,j}^{t},\tag{7}$$

where  $\binom{a}{b_1, b_2, \dots, b_m}$  denotes the number of pairwise disjoint subsets of sizes  $b_1, b_2, \dots, b_m$  respectively of a set of size a, and  $\lambda_{i,j}^t$  denotes the number of triples  $(X, Y, Z) \in C^3$  with  $|X\Delta Y| = i, |X\Delta Z| = j$ , and  $|(X\Delta Y) \cap (X\Delta Z)| = t$ , or equivalently, with  $|X\Delta Y| = i, |X\Delta Z| = j$ , and  $|Y\Delta Z| = i+j-2t$ . Set  $x_{i,j}^t = 0$  if  $\binom{n}{i-t,j-t,t} = 0$ .

The key part of Schrijver's semidefinite programming bound is that for each  $k = 0, 1, \dots, \lfloor \frac{n}{2} \rfloor$ , the matrices

$$\left(\sum_{t=0}^{n} \beta_{i,j,k}^{t} x_{i,j}^{t}\right)_{i,j=k}^{n-k} \tag{8}$$

and

$$\left(\sum_{t=0}^{n} \beta_{i,j,k}^{t} (x_{i+j-2t,0}^{0} - x_{i,j}^{t})\right)_{i,j=k}^{n-k}$$
(9)

are positive semidefinite, where  $\beta_{i,j,k}^t$  is given by

$$\beta_{i,j,k}^{t} = \sum_{u=0}^{n} (-1)^{u-t} {\binom{u}{t}} {\binom{n-2k}{u-k}} {\binom{n-k-u}{i-u}} {\binom{n-k-u}{j-u}}.$$
(10)

Since

$$|\mathcal{C}| = \sum_{i=0}^{n} \binom{n}{i} x_{i,0}^{0},\tag{11}$$

an upper bound on A(n,d) can be obtained by considering the  $x_{i,j}^t$  as variables and by

maximizing 
$$\sum_{i=0}^{n} {n \choose i} x_{i,0}^{0}$$
 (12)

subject to the matrices (8) and (9) are positive semidefinite for each  $k = 0, 1, ..., \lfloor \frac{n}{2} \rfloor$  and subject to the following conditions on the  $x_{i,j}^t$  (see [3]).

- (i)  $x_{0,0}^0 = 1$ .
- (ii)  $0 \le x_{i,j}^t \le x_{i,0}^0$  and  $x_{i,0}^0 + x_{j,0}^0 \le 1 + x_{i,j}^t$  for all  $i, j, t \in \{0, 1, \dots, n\}$ .
- (iii)  $x_{i,j}^t = x_{i',j'}^{t'}$  if (i',j',i'+j'-2t') is a permutation of (i,j,i+j-2t).

(iv) 
$$x_{i,j}^t = 0$$
 if  $\{i, j, i+j-2t\} \cap \{1, 2, \dots, d-1\} \neq \emptyset$ 

## C. Improved Schrijver's Semidefinite Programming Bound on A(n, d)

1) New Constraints for  $x_{i,j}^t$ : Let C be an (n,d) code and let  $x_{i,j}^t$  be defined by (7). Theorem 3: For all  $i, j, t \in \{0, 1, ..., n\}$  with  $\binom{n}{i-t, j-t, t} \neq 0$ ,

$$x_{i,j}^{t} \leq \frac{T(t,i,j-t,n-i,d)}{\binom{i}{t}\binom{n-i}{j-t}} x_{i,0}^{0}.$$
(13)

*Proof:* Recall that  $\lambda_{i,j}^t$  is the number of triples  $(X, Y, Z) \in C^3$  with  $|X\Delta Y| = i$ ,  $|X\Delta Z| = j$ , and  $|Y\Delta Z| = i + j - 2t$ . For any pair  $(X, Y) \in C^2$  with  $|X\Delta Y| = i$ , the number of  $Z \in C$  such that  $|Z\Delta X| = j$  and  $|Z\Delta Y| = i + j - 2t$  is upper bounded by  $A(n, \Lambda, d)$ , where  $\Lambda = \{(X, j), (Y, i + j - 2t)\}$ . By Proposition 1,

$$A(n,\Lambda,d) = T(t,i,j-t,n-i,d).$$
(14)

Since the number of pairs  $(X,Y)\in \mathcal{C}^2$  such that  $|X\Delta Y|=i$  is  $\lambda_{i,0}^0,$ 

$$\lambda_{i,j}^{t} \le T(t, i, j - t, n - i, d)\lambda_{i,0}^{0}.$$
(15)

Therefore,

$$\begin{aligned} x_{i,j}^t &= \frac{1}{|\mathcal{C}| \binom{n}{i-t,j-t,t}} \lambda_{i,j}^t \\ &\leq \frac{T(t,i,j-t,n-i,d)}{|\mathcal{C}| \binom{n}{i-t,j-t,t}} \lambda_{i,0}^0 \\ &= \frac{T(t,i,j-t,n-i,d) \binom{n}{i}}{\binom{n}{i-t,j-t,t}} x_{i,0}^0 \\ &= \frac{T(t,i,j-t,n-i,d)}{\binom{n}{i-t,j-t,t}} x_{i,0}^0. \end{aligned}$$

The following corollary was used in [3].

*Corollary 4:* For each  $j \in \{0, 1, ..., n\}$ ,

$$\binom{n}{j} x_{0,j}^0 \le A(n,d,j). \tag{16}$$

Proof: By Theorem 3, we have

$$x_{0,j}^{0} \leq \frac{T(0,0,j,n,d)}{\binom{0}{0}\binom{n}{j}} x_{0,0}^{0} = \frac{A(n,d,j)}{\binom{n}{j}}.$$
(17)

Remark 5: Theorem 3 improve the condition  $x_{i,j}^t \leq x_{i,0}^0$  in Schrijver's semidefinite programming bound since  $\frac{T(t,i,j-t,n-i,d)}{\binom{i}{i}\binom{n-i}{j-t}} \leq 1$  (in fact,  $\frac{T(t,i,j-t,n-i,d)}{\binom{i}{i}\binom{n-i}{j-t}}$  is much less than 1 in general). Similarly, Corollary 4 in many cases (of *i* and *j*) improve the condition  $x_{i,0}^0 + x_{j,0}^0 \leq 1 + x_{i,j}^t$  since  $x_{u,0}^0 = x_{0,u}^0 = \frac{A(n,d,u)}{\binom{n}{u}}$  is much less than  $\frac{1}{2}$  in general.

2) Delsarte's Linear Programming Bound and Its Improvements: Let C be an (n, d) code, the distance distribution  $\{B_i\}_{i=0}^n$  of C is defined by

$$B_{i} = \frac{1}{|\mathcal{C}|} \cdot |\{(X, Y) \in \mathcal{C}^{2} \mid |X\Delta Y| = i\}|.$$
(18)

By definition,

$$\binom{n}{i}x_{i,0}^0 = B_i \tag{19}$$

for each i = 0, 1, ..., n. Hence,  $\{\binom{n}{i} x_{i,0}^0\}_{i=0}^n$  is the distance distribution on C. The following result can be found for example in [7] or [6].

Theorem 6: (Delsarte's linear programming bound and its improvements). Let C be an (n, d) code with distance distribution  $\{B_i\}_{i=0}^n = \{\binom{n}{i} x_{i,0}^0\}_{i=0}^n$ . For k = 1, 2, ..., n,

$$\sum_{i=1}^{n} P_k(n;i)B_i \ge -\binom{n}{k}, \qquad (20)$$

where  $P_k(n; x)$  is the Krawtchouk polynomial given by

$$P_k(n;x) = \sum_{j=0}^k (-1)^j \binom{x}{j} \binom{n-x}{k-j}.$$
(21)

If  $M = |\mathcal{C}|$  is odd, then

$$\sum_{i=1}^{n} P_k(n;i)B_i \ge -\binom{n}{k} + \frac{1}{M}\binom{n}{k}.$$
(22)

If  $M = |\mathcal{C}| \equiv 2 \pmod{4}$ , then there exists  $t \in \{0, 1, \dots, n\}$  such that

$$\sum_{i=1}^{n} P_k(n;i)B_i \ge -\binom{n}{k} + \frac{2}{M} \left[\binom{n}{k} + P_k(n;t)\right].$$
(23)

3) Linear Constraints on Distance Distributions  $\{B_i\}_{i=0}^n$ : If some linear constraints are used to improve Delsarte's linear programming bound on A(n,d), then these constraints can still be added to Schrijver's semidefinite programming bound to improve upper bounds on A(n,d). The following constraints are due to Mounits, Etzion, and Litsyn (see [4, Theorems 9 and 10]).

Theorem 7: Let C be an (n, d) code with distance distribution  $\{B_i\}_{i=0}^n$ . Suppose that d is even and  $\delta = d/2$ . Then

$$B_{n-\delta} + \left\lfloor \frac{n}{\delta} \right\rfloor \sum_{i < \delta} B_{n-i} \le \left\lfloor \frac{n}{\delta} \right\rfloor$$
(24)

and

$$B_{n-\delta-i} + [A(n,d,\delta+i) - A(n-\delta+i,d,\delta+i)]B_{n-\delta+i} + A(n,d,\delta+i)\sum_{j>i}B_{n-\delta+j} \le A(n,d,\delta+i)$$
(25)

for all  $i = 1, 2, ..., \delta - 1$ .

Table I shows improved upper bounds on A(n, d) when linear constraints in Theorems 3, 6, and 7 are added to Schrijver's semidefinite programming bound (12). In the table, by Schrijver bound we mean upper bound obtained from Schrijver's semidefinite programming bound (12). Among improved upper bounds on A(n, d), there are two new upper bounds, namely

$$A(18,8) \le 71$$
 and  $A(19,8) \le 131$ .

The other best known upper bounds are from [8]. As in [3], all computations here were done by the algorithm SDPT3 available online on the NEOS Server for Optimization (http://www.neos-server.org/neos/solvers/index.html).

*Remark 8:* Since A(n,d) = A(n+1,d+1) if d is odd, we can always assume that d is even. If d is even, then A(n,d) is attained by a code with all codewords having even weights. Hence, in Schrijver's semidefinite programming bound, one can put  $x_{i,j}^t = 0$  if i or j is odd.

*Remark 9:* In Theorems 3 and 7, the values of A(n, d, w) and  $T(w_1, n_1, w_2, n_2, d)$  may have not yet been known. However, we can replace them by any of their upper bounds (see the proof of [4, Theorem 10] for the validity of this replacement in Theorem 7). While best known upper bounds on A(n, d, w) (which are mostly from [9], [5], [3], [10]) are used in our computations, all upper bounds on  $T(w_1, n_1, w_2, n_2, d)$  that we used are from the tables on Erik Agrell's website http://webfiles.portal.chalmers.se/s2/research/kit/bounds/dcw.html.

new	improved	
upper	Schijver	Schrijver
bound	bound	bound
71	71	80

TABLE I Improved upper bounds for  ${\cal A}(n,d)$ 

best upper

previously

bound

known

best

lower

bound

known

d

n

III. Upper bounds on A(n, d, w)

# A. Some Properties of A(n, d, w)

We begin with some elementary properties of A(n, d, w) which can be found in [2]. *Theorem 10:* 

$$A(n, d, w) = A(n, d+1, w),$$
 if d is odd, (26)

$$A(n, d, w) = A(n, d, n - w),$$
 (27)

$$A(n,2,w) = \binom{n}{w}, \qquad (28)$$

$$A(n, 2w, w) = \left\lfloor \frac{n}{w} \right\rfloor,\tag{29}$$

$$A(n, d, w) = 1, \quad \text{if } 2w < d.$$
 (30)

Remark 11: By (26) and (28), we can always assume that d is even and  $d \ge 4$ . Also, by (27), (29), and (30), we can assume that  $d < 2w \le n$ .

# B. Schrijver's Semidefinite Programming Bound on A(n, d, w)

Let C be an (n, d, w) constant-weight code and let v = n - w. For each t, s, i, and j, define

$$y_{i,j}^{t,s} = \frac{1}{|\mathcal{C}| \begin{pmatrix} w \\ i-t,j-t,t \end{pmatrix} \begin{pmatrix} v \\ i-s,j-s,s \end{pmatrix}} \mu_{i,j}^{t,s},$$
(31)

where  $\mu_{i,j}^{t,s}$  is the number of triples  $(X, Y, Z) \in \mathcal{C}^3$  with  $|X \setminus Y| = i, |X \setminus Z| = j, |(X \setminus Y) \cap (X \setminus Z)| = t$ , and  $|(Y \setminus X) \cap (Z \setminus X)| = s$ , or equivalently, with  $|X\Delta Y| = 2i, |X\Delta Z| = 2j, |Y\Delta Z| = 2(i + j - t - s)$ , and  $|X\Delta Y\Delta Z| = w + 2t - 2s$ . Set  $y_{i,j}^{t,s} = 0$  if either  $\binom{w}{i-t,j-t,t} = 0$  or  $\binom{v}{i-s,j-s,s} = 0$ .

In the previous section,  $\beta_{i,j,k}^t$  depends on *n*. Hence,  $\beta_{i,j,k}^t$  should be denoted by  $\beta_{i,j,k}^{t,n}$ . We will use the later notation in this section. As in [3], for each  $k = 0, 1, ..., \lfloor \frac{w}{2} \rfloor$  and each  $l = 0, 1, ..., \lfloor \frac{w}{2} \rfloor$ , the matrices

$$\left(\sum_{t,s} \beta_{i,j,k}^{t,w} \beta_{i,j,l}^{s,v} y_{i,j}^{t,s}\right)_{i,j \in W_k \cap V_l}$$
(32)

and

$$\left(\sum_{t,s} \beta_{i,j,k}^{t,w} \beta_{i,j,l}^{s,v} (y_{i+j-t-s,0}^{0,0} - y_{i,j}^{t,s})\right)_{i,j \in W_k \cap V_l}$$
(33)

are positive semidefinite, where  $W_k = \{k, k+1, \dots, w-k\}$  and  $V_l = \{l, l+1, \dots, v-l\}$ . Since

$$|\mathcal{C}| = \sum_{i=0}^{\min\{w,v\}} {w \choose i} {v \choose i} y_{i,0}^{0,0},$$
(34)

an upper bound on A(n, d, w) can be obtained by considering the  $y_{i,j}^{t,s}$  as variables and by

maximizing 
$$\sum_{i=0}^{\min\{w,v\}} {w \choose i} {v \choose i} y_{i,0}^{0,0}$$
(35)

subject to the matrices (32) and (33) are positive semidefinite for each  $k = 0, 1, ..., \lfloor \frac{w}{2} \rfloor$  and each  $l = 0, 1, ..., \lfloor \frac{w}{2} \rfloor$ , and subject to the following conditions.

(i)  $y_{0,0}^{0,0} = 1$ .

(ii) 
$$0 \le y_{i,j}^{t,s} \le y_{i,0}^{0,0}$$
 and  $y_{i,0}^{0,0} + y_{i,0}^{0,0} \le 1 + y_{i,j}^{t,s}$  for all  $i, j, t, s \in \{0, 1, \dots, \min\{w, v\}\}$ .

- (ii)  $y_{i,j}^{t,s} = y_{i',j'}^{t',s'}$  if t' s' = t s and (i', j', i' + j' t' s') is a permutation of (i, j, i + j t s).
- (iv)  $y_{i,j}^{t,s} = 0$  if  $\{2i, 2j, 2(i+j-t-s)\} \cap \{1, 2, \dots, d-1\} \neq \emptyset$ .

# C. Improved Schrijver's Semidefinite Programming Bound on A(n, d, w)

1) New Constraints for  $y_{i,j}^{t,s}$ : Let C be an (n, d, w) constant-weight code and let  $y_{i,j}^{t,s}$  be defined by (31). The following theorem corresponds to Theorem 3 in the previous section.

Theorem 12: For all  $i, j, s, t \in \{0, 1, \dots, \min\{w, v\}\}$  with  $\binom{w}{i-t, j-t, t} \neq 0$  and  $\binom{v}{i-s, j-s, s} \neq 0$ ,

$$y_{i,j}^{t,s} \le \frac{T(t,i,j-t,w-i,s,i,j-s,v-i,d)}{\binom{i}{t}\binom{w-i}{j-t}\binom{i}{s}\binom{v-i}{j-s}}y_{i,0}^{0,0}.$$
(36)

*Proof:* Suppose that  $(X, Y) \in C^2$  such that  $|X\Delta Y| = 2i$ . We claim that the number of codewords  $Z \in C$  such that  $|X\Delta Z| = 2j$ ,  $|Y\Delta Z| = 2(i + j - t - s)$ , and  $|X\Delta Y\Delta Z| = w + 2t - 2s$  is upper bounded by T(t, i, j - t, w - i, s, i, j - s, v - i, d). It is easy to see that this number is upper bounded by  $A(n, \Lambda, d)$ , where  $\Lambda = \{(0, w), (X, 2j), (Y, 2(i + j - t - s)), (X\Delta Y, w + 2t - 2s)\}$ . By Proposition 2,

$$A(n,\Lambda,d) = T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d),$$
(37)

where  $n_1 = n_3 = \frac{1}{2}|X\Delta Y| = i$ ,  $n_2 = d_1 - n_1 = w - i$ ,  $n_4 = n - i - (w - i) - i = v - i$ , and similarly,  $w_1 = i - t, w_2 = (w - i) - (j - t), w_3 = s, w_4 = j - s$ . Hence,

$$A(n, \Lambda, d) = T(i - t, i, (w - i) - (j - t), w - i, s, i, j - s, v - i, d)$$
  
=  $T(t, i, j - t, w - i, s, i, j - s, v - i, d),$  (38)

where the later equality comes from Proposition 22 (iii) in the appendix. Since the number of pairs  $(X, Y) \in C^2$ such that  $|X\Delta Y| = 2i$  is  $\mu_{i,0}^{0,0}$ ,

$$\mu_{i,j}^{t,s} \le T(t,i,j-t,w-i,s,i,j-s,v-i,d)\mu_{i,0}^{0,0}.$$
(39)

Therefore,

$$\begin{split} y_{i,j}^{t,s} &= \frac{1}{|\mathcal{C}| \begin{pmatrix} w \\ i-t,j-t,t \end{pmatrix} \begin{pmatrix} v \\ i-s,j-s,s \end{pmatrix}} \mu_{i,j}^{t,s} \\ &\leq \frac{T(t,i,j-t,w-i,s,i,j-s,v-i,d)}{|\mathcal{C}| \begin{pmatrix} w \\ i-t,j-t,t \end{pmatrix} \begin{pmatrix} v \\ i-s,j-s,s \end{pmatrix}} \mu_{i,0}^{0,0} \\ &= \frac{T(t,i,j-t,w-i,s,i,j-s,v-i,d)}{\begin{pmatrix} w \\ i-t,j-t,t \end{pmatrix} \begin{pmatrix} v \\ i-s,j-s,s \end{pmatrix} \begin{pmatrix} w \\ i \end{pmatrix}^{-1} \begin{pmatrix} v \\ i \end{pmatrix}^{-1} y_{i,0}^{0,0} \\ &= \frac{T(t,i,j-t,w-i,s,i,j-s,v-i,d)}{\begin{pmatrix} i \\ t \end{pmatrix} \begin{pmatrix} w \\ j-t \end{pmatrix} \begin{pmatrix} w \\ i \end{pmatrix} \begin{pmatrix} v \\ i \end{pmatrix}^{-1} \begin{pmatrix} v \\ i-s \end{pmatrix}} y_{i,0}^{0,0}. \end{split}$$

2) Delsarte's Linear Programming Bound: Let C be an (n, d, w) constant-weight code with distance distribution  $\{B_i\}_{i=0}^n$ . By definition of  $y_{i,j}^{t,s}$ ,

$$\binom{w}{i}\binom{v}{i}y_{i,0}^{0,0} = B_{2i} \tag{40}$$

for every i (note that  $B_0 = 1$  and  $B_i = 0$  whenever i is odd or 0 < i < d or i > 2w).

Theorem 13: (Delsarte's linear programming bound). If  $\{B_i\}_{i=0}^n$  is the distance distribution of an (n, d, w) constant-weight code, then for k = 1, 2, ..., w,

$$\sum_{i=d/2}^{w} q(k,i,n,w) B_{2i} \ge -1,$$
(41)

where

$$q(k, i, n, w) = \frac{\sum_{j=0}^{i} (-1)^{j} {\binom{k}{j}} {\binom{w-k}{i-j}} {\binom{n-w-k}{i-j}}}{{\binom{w}{i}} {\binom{n-w}{i}}}.$$
(42)

Specifying Delsarte's linear programming bound on A(n, d) gives the following linear constraints on  $B_i$ , which sometimes help reducing upper bounds on A(n, d, w) by 1 (see [6, Proposition 11]).

Theorem 14: Let C be an (n, d, w) constant-weight code with distance distribution  $\{B_i\}_{i=0}^n$ . For each k = 1, 2, ..., n,

$$\sum_{i=d/2}^{w} P_k^-(n;2i) B_{2i} \le \frac{2}{M} \Big[ \Big( \binom{n}{k} - r_k \Big) q_k (M - q_k) + r_k (q_k + 1) (M - q_k - 1) \Big], \tag{43}$$

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where  $q_k$  and  $r_k$  are the quotient and the remainder, respectively, when dividing  $MP_k^-(n;w)$  by  $\binom{n}{k}$ , i.e.

$$MP_k^-(n;w) = q_k \binom{n}{k} + r_k \tag{44}$$

with  $0 \le r_k < \binom{n}{k}$ , and where  $P_k^-(n;x)$  is defined by

$$P_k^-(n;x) = \sum_{\substack{j=0\\j \text{ odd}}}^n \binom{x}{j} \binom{n-x}{k-j}.$$
(45)

3) New Linear Constraints on Distance Distributions  $\{B_i\}_{i=0}^n$ : Linear constraints which correspond to those in Theorem 7 have not been studied for constant-weight codes even though similar constraints have been studied by Argrell, Vardy, and Zeger in [5] (see Theorem 21 below). We now present these constraints. Several new notations are needed. For convenience, we fix the following settings until the end of this section.

- C is an (n, d, w) constant-weight code with distance distribution  $\{B_i\}_{i=0}^n$  such that d is even and  $d < 2w \le n$ .
- Let v = n w. Since  $2w \le n, w \le v$ .
- Let  $H = \{d/2, d/2 + 1, \dots, w\}$ , which is the set of all positive integer *i* such that  $B_{2i}$  can be nonzero.
- For each i ∈ H, let V<sub>i</sub> be the set of all vectors X in F<sup>n</sup> such that X has exactly i ones on the first w coordinates and exactly i ones on the last v = n − w coordinates.
- For  $i \neq j$  both in H, define

$$m_{i,j} = \max\{d(X,Y) \mid X \in \mathcal{V}_i, Y \in \mathcal{V}_j\}.$$
(46)

• For each codeword X in C, let

$$S_{2i}(X) = \{ Y \in \mathcal{C} \mid d(X, Y) = 2i \},$$
(47)

which is the set of all codewords Y in C at distance 2i from X. By definition of  $\{B_i\}_{i=0}^n$ ,

$$B_{2i} = \frac{1}{|\mathcal{C}|} \sum_{X \in \mathcal{C}} |S_{2i}(X)|$$
(48)

for each  $i \in H$ .

• For each  $i \in H$ , let  $Q_i$  denote an integer such that

$$T(i, w, i, v, d) \le Q_i. \tag{49}$$

• For  $i \neq j$  both in H with  $i + j \geq v$  and  $m_{i,j} = d$ , let  $Q_{ji}$  denote an integer such that

$$T(w-j, i, v-j, i, d) \le Q_{ji},\tag{50}$$

*Proposition 15:* For  $i \neq j$  both in H,

$$m_{i,j} = a + b, \tag{51}$$

where

$$a = \begin{cases} i+j & \text{if } i+j < w\\ i+j-2(i+j-w) & \text{if } i+j \ge w \end{cases}$$

and

$$b = \begin{cases} i+j & \text{if } i+j < v\\ i+j-2(i+j-v) & \text{if } i+j \ge v \end{cases}$$

In particular, if  $i + j \ge v \ge w$ , then

$$m_{i,j} = 2(n-i-j).$$
 (52)

Proof: The proof is straightforward.

Lemma 16: For each  $i \in H$  and each codeword  $X \in C$ ,

$$|S_{2i}(X)| \le Q_i. \tag{53}$$

*Proof:* Let X be a codeword in C. It is easy to see that  $|S_{2i}(X)|$  is upper bounded by  $A(n, \Lambda, d)$ , where  $\Lambda = \{(0, w), (X, 2i)\}$ . By Propositions 1 and 22 (iii),

$$A(n,\Lambda,d) \le T(w-i,w,i,v,d) = T(i,w,i,v,d).$$
(54)

Hence,  $|S_{2i}(X)| \le T(i, w, i, v, d) \le Q_i$ .

Theorem 17: Suppose that  $H_1$  is a nonempty subset of H such that  $m_{i,j} < d$  for all  $i \neq j$  both in  $H_1$ . Then for each codeword  $X \in C$ ,  $S_{2i}(X)$  is nonempty for at most one i in  $H_1$ . Furthermore,

$$\sum_{i \in H_1} \frac{B_{2i}}{Q_i} \le 1.$$
(55)

*Proof:* Let X be a codeword in C. Suppose on the contrary that there exist  $i \neq j$  both in  $H_1$  such that  $S_{2i}(X)$ and  $S_{2j}(X)$  are nonempty. Then choose any  $Y \in S_{2i}(X)$  and  $Z \in S_{2j}(X)$ . By rearranging the coordinates, we may assume that

$$X = \underbrace{1\cdots 1}^{w} \underbrace{0\cdots 0}^{v}.$$
(56)

Since d(X,Y) = 2i and X and Y have the same weight w, Y + X must have exactly i ones on the first w coordinates and exactly i ones on the last v coordinates. This means  $Y + X \in \mathcal{V}_i$ . Similarly,  $Z + X \in \mathcal{V}_j$ . By definition of  $m_{i,j}, d(Y + X, Z + X) \leq m_{i,j}$ . Thus,

$$d(Y,Z) = d(Y+X,Z+X) \le m_{i,j} < d,$$
(57)

which is a contradiction since Y and Z are two different codewords in C. Hence,  $S_{2i}(X)$  is nonempty for at most one i in  $H_1$ . It follows by Lemma 16 that

$$\sum_{i \in H_1} \frac{|S_{2i}(X)|}{Q_i} \le 1.$$
(58)

Taking sum of (58) over all  $X \in C$ , we get

$$\sum_{i \in H_1} \frac{B_{2i}}{Q_i} \le 1.$$
(59)

We now consider the case  $m_{i,j} = d$  for some  $i \neq j$  both in H. The following Lemma says that the existence of a codeword at distance 2i from X may reduce the total number of codewords at distance 2j from X.

Lemma 18: Suppose  $i \neq j$  both in H such that  $i + j \geq v$  and  $m_{i,j} = d$ . If X is a codeword in C such that  $|S_{2i}(X)| \geq 1$ , then

$$|S_{2j}(X)| \le Q_{ji}.\tag{60}$$

*Proof:* Fix a codeword  $Y \in S_{2i}(X)$ . If  $S_{2j}(X)$  is empty, then there is nothing to prove. Hence, we assume  $|S_{2j}(X)| \ge 1$ . Let  $Z \in S_{2j}(X)$ . By rearranging the coordinates, we may assume that

$$X = \underbrace{1\cdots 1}^{w} \underbrace{0\cdots 0}^{v} \tag{61}$$

As in the proof of Theorem 17, we can show that  $Y + X \in \mathcal{V}_i$  and  $Z + X \in \mathcal{V}_j$ . By definition of  $m_{i,j}$ ,

$$d \le d(Y, Z) = d(Y + X, Z + X) \le m_{i,j} = d.$$
(62)

Thus,

$$d(Y,Z) = d(Y+X,Z+X) = m_{i,j} = d.$$
(63)

Since  $i + j \ge v \ge w$ , by rearranging the first w coordinates, we may assume that on the first w coordinates:

$$Y + X = 1 \cdots 1 \quad 1 \cdots 1 \quad \overbrace{0 \cdots 0}^{w-i} | \cdots$$
  

$$Z + X = \underbrace{0 \cdots 0}_{w-j} \quad \underbrace{1 \cdots 1}_{i+j-w} \quad 1 \cdots 1 | \cdots$$
(64)

On the first w coordinates, Z + X must have exactly i + j - w ones on the first i coordinates (the other w - i ones of Z + X must be fixed since  $d(Y + X, Z + X) = m_{i,j}$ ).

Similarly, since  $i + j \ge v$ , by rearranging the last v coordinates, we may assume that on the last v coordinates:

$$Y + X = \cdots | 1 \cdots 1 1 \cdots 1 0 \cdots 0$$
  

$$Z + X = \cdots | 0 \cdots 0 \underbrace{1 \cdots 1}_{v-j} 1 \cdots 1 \cdots 1$$
(65)

On the last v coordinates, Z + X must have exactly i + j - v ones on the first i coordinates (the other v - i ones of Z + X must be fixed since  $d(Y + X, Z + X) = m_{i,j}$ ).

From (61), (64), and (65), we get

$$d(Z, X + Y) = wt(X + Y + Z)$$
  
=  $wt(X + (Y + X) + (Z + X))$   
=  $(i + j - w) + (v - j + v - i)$   
=  $2v - w.$  (66)

Now the number of  $Z \in S_{2j}(X)$  is upper bounded by  $A(n, \Lambda, d)$ , where  $\Lambda = \{(0, w), (X, 2j), (Y, d), (X + Y, 2v - w)\}$ . By Proposition 15,

$$d = m_{i,j} = 2(n - i - j).$$
(67)

Applying Proposition 2, we get (by replacing d = 2(n - i - j) and n = w + v)

$$A(n, \Lambda, d) = T(w - j, i, 0, w - i, i + j - v, i, v - i, v - i, d)$$
  
= T(w - j, i, v - j, i, d), (68)

where the last equality comes from Proposition 22 in the appendix. Therefore,

$$S_{2j}(X)| \leq A(n, \Lambda, d)$$
  
=  $T(w - j, i, v - j, i, d)$   
 $\leq Q_{ji}.$  (69)

Theorem 19: Suppose that  $H_1$  is a subset of H satisfying the following properties.

- $|H_1| \ge 2.$
- There exist  $i \neq j$  both in  $H_1$  such that  $i + j \ge v$  and  $m_{i,j} = d$ .
- For all  $k \neq l$  both in  $H_1$  such that either  $k \notin \{i, j\}$  or  $l \notin \{i, j\}$ , we always have  $m_{k,l} < d$ .

Let  $H_2 = H_1 \setminus \{i, j\}$ . Then

$$\frac{Q_j - Q_{ji}}{Q_j Q_{ij}} B_{2i} + \frac{1}{Q_j} B_{2j} + \sum_{k \in H_2} \frac{1}{Q_k} B_{2k} \le 1, \quad \text{if } \frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \ge 1,$$
(70)

$$\frac{1}{Q_i}B_{2i} + \frac{Q_i - Q_{ij}}{Q_i Q_{ji}}B_{2j} + \sum_{k \in H_2} \frac{1}{Q_k}B_{2k} \le 1, \quad \text{if } \frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \ge 1,$$
(71)

$$\sum_{k \in H_1} \frac{1}{Q_k} B_{2k} \le 1, \quad \text{if } \frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \le 1.$$
(72)

*Proof:* We first prove (70). It suffices to show that for every codeword X in C,

$$\frac{Q_j - Q_{ji}}{Q_j Q_{ij}} |S_{2i}(X)| + \frac{1}{Q_j} |S_{2j}(X)| + \sum_{k \in H_2} \frac{1}{Q_k} |S_{2k}(X)| \le 1,$$
(73)

if  $\frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \ge 1$ . Let X be any codeword in C. By Lemma 16,

$$|S_{2i}(X)| \le Q_i \quad \text{and} \quad |S_{2j}(X)| \le Q_j.$$
(74)

By Lemma 18,

$$|S_{2i}(X)| \le Q_{ij} \text{ if } |S_{2j}(X)| \ge 1,$$
(75)

$$|S_{2j}(X)| \le Q_{ji} \text{ if } |S_{2i}(X)| \ge 1.$$
(76)

We prove (73) by considering the following three cases.

Case 2:  $|S_{2i}(X)| \ge 1$  and  $|S_{2j}(X)| = 0$ . Since  $|S_{2i}(X)| \ge 1$ ,  $|S_{2k}(X)| = 0$  for every  $k \in H_2$  by Theorem 17. Hence, to prove (73), we only need to prove that

$$(Q_j - Q_{ji})|S_{2i}(X)| \le Q_j Q_{ij}.$$
(77)

By hypothesis,  $\frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \ge 1$ . Thus,  $(Q_j - Q_{ji})Q_i \le Q_jQ_{ij}$  and hence

$$(Q_j - Q_{ji})|S_{2i}(X)| \le (Q_j - Q_{ji})Q_i \le Q_j Q_{ij}.$$
(78)

Case 3:  $|S_{2i}(c)| \ge 1$  and  $|S_{2j}(c)| \ge 1$ . As in Case 2,  $|S_{2k}(X)| = 0$  for every  $k \in H_2$ . We have

$$\frac{Q_{j} - Q_{ji}}{Q_{j}Q_{ij}} |S_{2i}(X)| + \frac{1}{Q_{j}} |S_{2j}(X)| \leq \frac{Q_{j} - Q_{ji}}{Q_{j}Q_{ij}} Q_{ij} + \frac{1}{Q_{j}} Q_{ji} \\
= 1 - \frac{Q_{ji}}{Q_{j}} + \frac{Q_{ji}}{Q_{j}} \\
= 1.$$
(79)

Therefore, (73) is proved and so is (70).

By symmetry, (71) follows.

We now prove (72). It suffices to show that for every codeword X in C,

$$\sum_{k \in H_1} \frac{1}{Q_k} |S_{2k}(X)| \le 1,$$
(80)

if  $\frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} \le 1$ . If either  $|S_{2i}(X)| = 0$  or  $|S_{2j}(X)| = 0$ , then proving (80) is exactly the same as proving (58). Hence, suppose that  $|S_{2i}(X)| \ge 1$  and  $|S_{2j}(X)| \ge 1$ . As in Case 2,  $|S_{2k}(X)| = 0$  for every  $k \in H_2$ . We have

$$\frac{1}{Q_i}|S_{2i}(X)| + \frac{1}{Q_j}|S_{2j}(X)| \le \frac{1}{Q_i}Q_{ij} + \frac{1}{Q_j}Q_{ji} \le 1.$$
(81)

We now specify which  $H_1$  are used in Theorems 17 and 19. Let

$$\alpha = d/2 - (n - 2w) \tag{82}$$

and let

$$\alpha_1 = \left\lfloor \frac{\alpha+1}{2} \right\rfloor \text{ and } \alpha_2 = \left\lfloor \frac{\alpha}{2} \right\rfloor$$
(83)

so that  $\alpha_1 + \alpha_2 = \alpha$ . Also, let

$$i_0 = w - \alpha_1 \text{ and } j_0 = w - \alpha_2.$$
 (84)

• Case 1:  $\alpha$  is even. In this case,  $i_0 = j_0$ . We apply Theorem 17 for

$$H_1 = \{j_0, j_0 + 1, \dots, w\}$$
(85)

and apply Theorem 19 for

$$H_1 = \{i_0 - \epsilon, j_0 + \epsilon, j_0 + \epsilon + 1, \dots, w\}$$
(86)

(with  $i = i_0 - \epsilon$  and  $j = j_0 + \epsilon$ ) for each  $\epsilon = 1, 2, \cdots, w - j_0$ .

• Case 2:  $\alpha$  is odd. In this case,  $i_0 < j_0$ . We apply Theorem 19 for

$$H_1 = \{i_0 - \epsilon, j_0 + \epsilon, j_0 + \epsilon + 1, \dots, w\}$$
(87)

(with  $i = i_0 - \epsilon$  and  $j = j_0 + \epsilon$ ) for each  $\epsilon = 0, 1, \dots, w - j_0$ .

*Example 20:* Consider (n, d, w) = (27, 8, 13). We have  $\alpha = d/2 - (n - 2w) = 3$  is odd. Hence,  $\alpha_1 = 2$  and  $\alpha_2 = 1$ . So,  $i_0 = 11$  and  $j_0 = 12$ . We can apply Theorem 19 for  $H_1 = \{i = i_0, j = j_0, w\} = \{11, 12, 13\}$  (with  $\epsilon = 0$ ). We have

$$\begin{aligned} Q_i &= 26 \geq T(2, 13, 3, 14, 8) = T(11, 13, 11, 14, 8), \\ Q_j &= 1 = T(1, 13, 2, 14, 8) = T(12, 13, 12, 14, 8), \\ Q_{ij} &= 20 \geq T(2, 12, 3, 12, 8), \\ Q_{ji} &= 1 = T(1, 11, 2, 11, 8), \end{aligned}$$

and

 $Q_k = 1 = T(0, 13, 1, 14, 8) = T(13, 13, 13, 14, 8)$ 

for k = 13. Since  $\frac{Q_{ij}}{Q_i} + \frac{Q_{ji}}{Q_j} = \frac{20}{26} + \frac{1}{1} \ge 1$ , Theorem 19 gives

$$B_{24} + B_{26} \le 1 \tag{88}$$

and

$$\frac{1}{26}B_{22} + \frac{26 - 20}{26}B_{24} + B_{26} \le 1.$$
(89)

The later constraint is equivalent to

$$B_{22} + 6B_{24} + 26B_{26} \le 26. \tag{90}$$

For  $H_1 = \{10, 13\}$  (with  $\epsilon = 1$ ), Theorem 19 gives less effective linear constraints.

When  $\alpha \leq 0$ , there is no set  $H_1$  satisfying Theorem 19. In this case, the following type of linear constraints which comes from [5, Proposition 17] is very useful. As in [5], let  $T'(w_1, n_1, w_2, n_2, d)$  be the largest possible size of a  $(w_1, n_1, w_2, n_2, d)$  doubly-bounded-weight code (a  $(w_1, n_1, w_2, n_2, d)$  doubly-bounded-weight code is an  $(n_1 + n_2, d, w_1 + w_2)$  constant-weight code such that every codeword has at most  $w_1$  ones on the first  $n_1$  coordinates). Tables for upper bounds on  $T'(w_1, n_1, w_2, n_2, d)$  can be found on Erik Agrell's website http://webfiles.portal.chalmers.se/s2/research/kit/bounds/dbw.html.

Theorem 21: Let  $\delta = d/2$ . For  $i, j \in \{\delta, \delta + 1, \dots, w\}$  with  $i \neq j$ . If  $i + j \leq n - \delta$ , define  $P_{ij}$  and  $P_{ji}$  as any nonnegative integers such that

 $P_{ij} \ge \min\{P_i, T'(\Delta, j, i - \Delta, n - w - j, 2i - 2\Delta\},\tag{91}$ 

$$P_{ji} \ge \min\{P_j, T'(\Delta, i, j - \Delta, n - w - i, 2j - 2\Delta\},\tag{92}$$

			best	best upper		
			lower	bound	new	
			bound	previously	upper	Schrijver
n	d	w	known	known	bound	bound
20	6	8	588	1107	1106	1136
22	8	10	616	634	630	634
23	8	9	400	707	703	707
26	8	9	887	2108	2104	2108
26	8	11	1988	5225	5208	5225
27	8	9	1023	2914	2882	2918
27	8	11	2404	7833	7754	7833
27	8	12	3335	10547	10460	10697
27	8	13	4094	11981	11897	11981
28	8	9	1333	3895	3886	3900
28	8	11	3773	11939	11896	12025
28	8	12	4927	17011	17008	17011
28	8	13	6848	21152	21148	21152
23	10	9	45	81	79	82
25	10	11	125	380	379	380
25	10	12	137	434	433	434
26	10	11	168	566	565	566
26	10	12	208	702	691	702
27	10	11	243	882	871	882
27	10	12	351	1201	1190	1201
27	10	13	405	1419	1406	1419
28	10	11	308	1356	1351	1356
25	12	10	28	37	36	37

TABLE II New upper bounds for  ${\cal A}(n,d,w)$ 

where  $\Delta := w - \delta$ . Also, define  $P_k := Q_k$  for each  $k \in H$ . Then

$$P_{ji}B_{2i} + (P_i - P_{ij})B_{2j} \le P_i P_{ji}, \quad \text{if } \frac{P_{ij}}{P_i} + \frac{P_{ji}}{P_j} > 1,$$
(93)

$$(P_j - P_{ji})B_{2i} + P_{ij}B_{2j} \le P_j P_{ij}, \quad \text{if } \frac{P_{ij}}{P_i} + \frac{P_{ji}}{P_j} > 1,$$
(94)

$$P_j B_{2i} + P_i B_{2j} \le P_i P_j, \quad \text{if } \frac{P_{ij}}{P_i} + \frac{P_{ji}}{P_j} \le 1.$$
 (95)

### APPENDIX

## UPPER BOUNDS ON $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$

To apply Theorem 12, we need tables of upper bounds on  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$ . However, there are no such tables available since this is the first time the function  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$  is introduced. We show here some elementary properties that are used to obtain upper bounds on  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$ .

In general, let us define  $T(\{(w_i, n_i)\}_{i=1}^t, d)$  as follows. For  $t \ge 1$ , a  $(\{(w_i, n_i)\}_{i=1}^t, d)$  multiply constant-weight code is a  $(\sum_{i=1}^t n_i, d)$  code such that there are exactly  $w_i$  ones on the  $n_i$  coordinates. When t = 1 this is definition of an  $(n_1, d, w_1)$  constant-weight code, when t = 2 this is definition of a  $(w_1, n_1, w_2, n_2, d)$  doubly-constant-weight code, etc.. Let  $T(\{(w_i, n_i)\}_{i=1}^t, d)$  be the largest possible size of a  $(\{(w_i, n_i)\}_{i=1}^t, d)$  multiply constant-weight code.

We present here elementary properties that are used to get upper bounds on  $T(\{(w_i, n_i)\}_{i=1}^t, d)$ . The proofs of these properties are similar to those for A(n, d, w) or  $T(w_1, n_1, w_2, n_2, d)$ , and hence are omitted. Upper bounds on  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$  that we used in Theorem 12 are the best upper bounds obtained from these properties.

Proposition 22: (i) If d is odd then,

$$T(\{(w_i, n_i)\}_{i=1}^t, d) = T(\{(w_i, n_i)\}_{i=1}^t, d+1).$$
(96)

(ii) If  $w_j = 0$  for some  $j \in \{1, 2, ..., t\}$ , then

$$T(\{(w_i, n_i)\}_{i=1}^t, d) = T(\{(w_i, n_i)\}_{i \neq j}, d).$$
(97)

- (iii)  $T(\{(w_i, n_i)\}_{i=1}^t, d)$  does not change if we replace any  $w_i$  by  $n_i w_i$ .
- (iv)  $T(\{(w_i, n_i)\}_{i=1}^t, 2) = \prod_{i=1}^t \binom{n_i}{w_i}.$
- (v)  $T(\{(w_i, n_i)\}_{i=1}^t, 2\sum_{i=1}^t w_i) = \min_{1 \le i \le t} \left\lfloor \frac{n_i}{w_i} \right\rfloor.$
- (vi)  $T(\{(w_i, n_i)\}_{i=1}^t, d) = 1$  if  $2\sum_{i=1}^t w_i < d$ .

*Remark 23:* By (i) and (iv), we can always assume that d is even and  $d \ge 4$ . By (ii) and (iii), we may assume that  $0 < 2w_i \le n_i$  for each i. Also, by (v) and (vi), we can assume that  $d < 2\sum_{i=1}^{t} w_i$ .

The next proposition can be used to reduce the size of  $\{(w_i, n_i)\}_{i=1}^t$  from t to t-1. When the size of the set is 2, we use known upper bounds on  $T(w_1, n_1, w_2, n_2, d)$ .

Proposition 24: If  $t \ge 2$ , then

$$T(\{(w_i, n_i)\}_{i=1}^t, d) \le T(\{(w'_i, n'_i)\}_{i=1}^{t-1}, d),$$
(98)

where  $w'_i = w_i, n'_i = n_i$  for i = 1, 2, ..., t - 2, and  $w'_{t-1} = w_{t-1} + w_t, n'_{t-1} = n_{t-1} + n_t$ .

Proposition 25: If  $w_i > 0$ , then

$$T(\{(w_i, n_i)\}_{i=1}^t, d) \le \left\lfloor \frac{n_i}{w_i} T(\{(w'_i, n'_i)\}_{i=1}^t, d) \right\rfloor,$$
(99)

where  $\{(w'_i, n'_i)\}_{i=1}^t$  is obtained from  $\{(w_i, n_i)\}_{i=1}^t$  by replacing the pair  $(w_i, n_i)$  by  $(w_i - 1, n_i - 1)$ .

Proposition 26: If  $w_i < n_i$ , then

$$T(\{(w_i, n_i)\}_{i=1}^t, d) \le \left\lfloor \frac{n_i}{n_i - w_i} T(\{(w_i', n_i')\}_{i=1}^t, d) \right\rfloor,$$
(100)

where  $\{(w'_i, n'_i)\}_{i=1}^t$  is obtained from  $\{(w_i, n_i)\}_{i=1}^t$  by replacing the pair  $(w_i, n_i)$  by  $(w_i, n_i - 1)$ .

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