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# Abstract

#### Abstract:

Let A(n,d) (resp. A(n,d,w)) be the maximal cardinality of codes (resp. constant-weight codes of weight w) of length n and minimum distance d. We introduce our recent results on A(n,d) and A(n,d,w). We improve two values of A(n,d) and twenty one values of A(n,d,w).



- I. Coding theory
- II. Improvements on optimal codes
- III. Improvements on optimal constant-weight codes

# I. Coding Theory

# Coding Theory

Let  $B = \{0, 1\}$  be the set of binary alphabets. In coding theory, one would like to study the set  $X = B^n$  of n-tuples of alphabets. We can view X from various point of views:

- (a) X is a set.
- (b) X is an abelian group when we give an abelian group structure on B.
- (c) X is a vector space when we give a field structure on B. Sometimes we denoted this field by  $\mathbb{F}_2$  and the *n*-dimensional vector space over  $\mathbb{F}_2$  by  $\mathbb{F}_2^n$ .

#### Coding Theory-Cont.

- (d)  $\mathbb{F}_2^n$  is a metric space when it is equipped with the Hamming metric:
  - For  $x = (x_1, \dots, x_n)$  and  $y = (y_1, \dots, y_n)$  in  $\mathbb{F}_2^n$ , we define  $d(x, y) = |\{i \mid x_i \neq y_i\}|.$

Sometimes this metric space is denoted by  $(\mathbb{F}_2^n, d_H)$ .

- (e) X is a graph where two vectors  $x, y \in X$ , x and y are adjacent iff  $d_H(x, y) = 1$ . This graph is called the Hamming graph, and it is a distance regular graph.
- (f) X is an association scheme which is called the Hamming scheme.
- (g) X form an affine geometry AG(n,2) of dimension n over  $\mathbb{F}_2$ .

### What is a code?

Binary code:

- A subset  $\mathcal{C}$  of  $B^n$  is called a *(binary) code* of length n.
- An element of a code  $\mathcal{C}$  is called a *codeword*.

Minimum distance of a code:

• For 
$$X = (x_1, \ldots, x_n)$$
 and  $Y = (y_1, \ldots, y_n)$  in  $B^n$ , define

 $d(X, Y) = |\{i \mid x_i \neq y_i\}|.$ 

• Minimum distance of a code  $\mathcal{C}$  is defined by

 $\min\{d(X,Y)|X,Y\in\mathcal{C},X\neq Y\}.$ 

Codes can correct errors

Three basic parameters of a code: length n, cardinality  $|\mathcal{C}|$ , minimum distance d. The minimum distance determine the error capability:

**Theorem:** A code with minimum distance d can correct up to  $\lfloor \frac{d-1}{2} \rfloor$  errors.

## A fundamental problem on coding theory

A natural problem:

Given n, find a code of length n having:

- large minimum distance,
- large number of codewords.

**Remark:** We can't get both.

### A fundamental problem on coding theory-Cont.

**Definition:** Given n and d, define

$$A(n,d) =$$
 maximum number of codewords  
in any code of length  $n$  and  
minimum distance  $> d$ .

Elementary properties of A(n,d)

Elementary properties of A(n,d):

(1) 
$$A(n, 2d) = A(n - 1, 2d - 1),$$

(2) 
$$A(n,d) \le 2A(n-1,d).$$

These properties are useful when n or d is small.

Some upper bounds for A(n,d)

To investigate A(n,d) for larger values of n and d, we need some theory for upper bounds on A(n,d). The upper bounds which will be useful in our investigation are:

- Hamming bound
- Johnson bound
- Delsarte's linear programming bounds
- Schrijver's semi-definite programming bounds

### Hamming bounds

Theorem (Hamming):

$$A(n, 2d+1) \le \frac{2^n}{1 + \binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{d}}.$$

**Quick reason:** The balls of radius *d* centered at codewords are mutually disjoint!

### Johnson bound

We need the concept of 'optimal constant-weight codes' in introducing Johnson bound:

**Definition:** Given n, d, and w, define

A(n, d, w) = maximum number of codewords in any code of length n and minimum distance  $\geq d$  such that each codeword has exactly w ones.

**Remark:** Codes in the above definition are called *constant-weight codes* of length n and weight w.

### Johnson bound-Cont.

Theorem (Johnson):

$$A(n, 2d+1) \le \frac{2^n}{1 + \binom{n}{1} + \dots + \binom{n}{d} + \frac{\binom{n}{d+1} - \binom{2d+1}{d}A(n, 2d+2, 2d+1)}{\left[\frac{n}{d+1}\right]}}$$

Quick reason: Hamming bound + careful consideration on spheres of radius d + 1 centered at codewords. Why A(n, 2d + 2, 2d + 1) appears in this consideration? Elementary properties of A(n, d, w)

Elementary properties of A(n, d, w): (1) A(n, d, w) = A(n, d+1, w), if d is odd, (2) A(n, d, w) = A(n, d, n - w),(3)  $A(n, 2, w) = \binom{n}{w},$ (4)  $A(n, 2w, w) = \lfloor \frac{n}{w} \rfloor$ , (5) A(n, d, w) = 1, if 2w < d, (6)  $A(n,d,w) \leq \lfloor \frac{n}{w}A(n-1,d,w-1) \rfloor$ , (7)  $A(n,d,w) \leq \lfloor \frac{n}{n-w} A(n-1,d,w) \rfloor.$ 

### Philosophy involved in Johnson's result

If we have a 'good' upper bound for A(n, d, w), then it would induce a 'good' upper bound for A(n, d). In this point of view, we need to develop 'coding theory' on  $Y = B^{n,w}$  where  $B^{n,w}$  denotes the set of binary *n*-vectors of weight *w*. We first trying to understand *Y* from various point of views:

- (1) Y is a set.
- (2) Y is equipped with a metric, since it is a subset of a metric space.
- (3) Y is a graph which is called the Johnson graph.
- (4) Y is an association scheme which is called the Johnson scheme.

### Philosophy involved in Johnson's result-Cont.

- (1) Since Y is a set equipped with a metric, Hamming type theorem can be developed for A(n, d, w). In this case, the size of a ball of radius 2r is:  $1 + {w \choose 1} {n-w \choose 1} + \dots + {w \choose r} {n-w \choose r}$ .
- (2) We can also develop Johnson type theorem for A(n, d, w) by considering Hamming type theorem + careful consideration on spheres of radius t in Y.
- (3) This means that we need to consider a subset of vectors in X which is at distance r from one point, and at distance t from another point.

A general definition for A(n,d), A(n,d,w)

**Definition:** For a finite (possibly empty) set  $\Lambda = \{(X_i, d_i)\}_{i \in I}$ , where each  $X_i$  is a vector in X and each  $d_i$  is a nonnegative integer, we define

> $A(n, \Lambda, d) =$  maximum number of codewords in any binary code of length nand minimum distance d such that each codeword is at distance  $d_i$ from  $X_i$  for all  $i \in I$ .

### A general definition-Cont.

**Special case 1:**  $|\Lambda| = 0$  We get the usual definition of A(n, d).

**Special case 2:**  $|\Lambda| = 1$  Suppose  $\Lambda = \{(X_1, d_1)\}$ . By translation, we may assume that  $X_1$  is the zero vector. Hence,

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

where  $w = d_1$ .

What will happen if  $|\Lambda| = 2$ ?

Doubly-constant-weight codes

#### **Definition:**

 $T(w_1, n_1, w_2, n_2, d) =$ maximum number of codewords in any code of length n and minimum distance  $\geq d$  such that each codeword has exactly  $w_1$  ones on the first  $n_1$  coordinates and exactly  $w_2$  ones on the last  $n_2$  coordinates.

**Remark:** Codes in the above definition of  $T(w_1, n_1, w_2, n_2, d)$  are called *doubly-constant-weight codes*.

#### A general definition-Cont.

**Special case 3:**  $|\Lambda| = 2$  Let  $\Lambda = \{(X_1, d_1), (X_2, d_2)\}$ . We have the following proposition.

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

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

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  $w_2 = \frac{1}{2}(d_1 + d_2 - n_1)$ .

#### Delsarte's Linear Programming bounds

#### **Distance distribution**:

- Let  $\mathcal{C}$  be a code (of length n).
- For each  $i = 0, 1, \ldots, n$ , define

$$B_i = \frac{1}{|\mathcal{C}|} \cdot |\{(X, Y) \in \mathcal{C}^2 | d(X, Y) = i\}|.$$

• The set  $\{B_i\}_{i=0}^n$  is called the *distance distribution* of  $\mathcal{C}$ .

Important equality:

$$B_0 + B_1 + \dots + B_n = |\mathcal{C}|.$$

#### Delsarte's Theorem

**Theorem (Delsarte, 1973):** Let C be a code with distance distribution  $\{B_i\}_{i=0}^n$ . Then

$$\sum_{i=0}^{n} P_k(n;i)B_i \ge 0$$

for each k = 1, 2, ..., n, where  $P_k(n; x)$  is the Krawtchouk polynomial given by

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

### Delsarte linear programming (LP) bound

**Delsarte LP bound** Consider  $B_i$ 's as variables. Then

$$A(n,d) \le 1 + \max \lfloor B_1 + \dots + B_n \rfloor,$$

where the maximization is taken over all  $(B_1, \ldots, B_n)$  satisfying the linear constraints of Delsarte's theorem and satisfying  $B_i \ge 0$ for  $i = 1, \ldots, n$ .

### Schrijver SDP bound

Definition (Triple distance distribution):

• Let  $\mathcal{C}$  be a code.

• For each 
$$i, j, t \in \{0, 1, \dots, n\}$$
, define  

$$\lambda_{i,j}^t = \left| \begin{cases} (X, Y, Z) \in \mathcal{C}^3 \\ (X, Y, Z) \in \mathcal{C}^3 \\ d(Y, Z) = i, d(X, Z) = j, \\ d(Y, Z) = i + j - 2t. \end{cases} \right\}$$
• Define  $\binom{n}{a,b,c} = \frac{n!}{a!b!c!(n-a-b-c)!}$ .  
• If  $\binom{n}{i-t,j-t,t} \neq 0$ , let  
 $x_{i,j}^t = \frac{1}{|\mathcal{C}|\binom{n}{i-t,j-t,t}} \lambda_{i,j}^t$ .  
• If  $\binom{n}{i-t,j-t,t} = 0$ , let  $x_{i,j}^t = 0$ .

## Schrijver SDP bound-Cont.

#### **Remark:**

• For each 
$$i = 0, 1, ..., n$$
,

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

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

### Schrijver's result

**Theorem (Schrijver, 2005)** For  $k = 0, 1, \ldots, \lfloor \frac{n}{2} \rfloor$ , the matrices

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

and

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

are positive semidefinite, where

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

#### Schrijver's result-Cont.

Idea of the proof (1):

- Let  $\mathcal{P}$  be the collection of all subsets of  $\{1, 2, \ldots, n\}$  (which can be identified with  $\mathcal{F}^n$ ).
- For i, j, t, let  $M_{i,j}^t$  be the  $\mathcal{P} \times \mathcal{P}$  matrix with  $(M_{i,j}^t)_{X,Y} = \begin{cases} 1 & \text{if } |X| = i, |Y| = j, |X \cap Y| = t \\ 0 & \text{otherwise} \end{cases}$
- Let

$$\mathcal{A}_n = \left\{ \sum_{i,j,t=0}^n c_{i,j}^t M_{i,j}^t \mid c_{i,j}^t \in \mathbb{C} \right\}.$$

## Schrijver's result

Idea of the proof (2):

- $\mathcal{A}_n$  is called the Terwilliger algebra of the Hamming cube  $\mathcal{P} \equiv \mathcal{F}^n$ .
- There exists a unitary matrix U such that

$$U^* \mathcal{A}_n U = \left\{ \begin{pmatrix} C_0 & 0 & \cdots & 0 \\ 0 & C_1 & \cdots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \cdots & C_m \end{pmatrix} \right\},\$$

where each  $C_k$  runs over all matrices of the form

$$\left( egin{array}{ccccc} B_k & 0 & \cdots & 0 \\ 0 & B_k & \cdots & 0 \\ dots & dots & \ddots & 0 \\ 0 & 0 & \cdots & B_k \end{array} 
ight)$$

• Choosing 2 (suitable) positive semidefinite elements R, R' of  $\mathcal{A}_n$ , Schrijver get the desired result.

Schrijver semidefinite programming (SDP) bound

#### Schrijver SDP bound:

$$A(n,d) \le \max \sum_{i=0}^{n} \binom{n}{i} x_{i,0}^{0}$$

subject to the matrices in the above Schrijver's Theorem are positive semidefinite for each  $k = 0, 1, \ldots, \lfloor \frac{n}{2} \rfloor$  and subject to the following conditions on  $x_{i,j}^t$ .

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

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

### II. Improvements on optimal codes

#### Some improvements on LP bound

Theorem (Mounits, Etzion, and Litsyn, 2002) Suppose C is a code of length n and minimum distance d. Let  $\delta = d/2$ . Then

• 
$$B_i \leq A(n, d, i)$$
 for  $i = 1, \dots, n$ ,

• 
$$B_{n-\delta} + \left\lfloor \frac{n}{\delta} \right\rfloor \sum_{i < \delta} B_{n-i} \le \left\lfloor \frac{n}{\delta} \right\rfloor,$$

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

for each  $i, 0 < i < \delta$ .

Key point of improvements of LP bound

#### In Delsarte LP bound

- We deal with the distance distribution  $\{B_i\}_{i=0}^n$ .
- When counting pairs  $(X, Y) \in C^2$  such that d(X, Y) = i, we can first fix X and count Y (and then take sum over all X).
- This means we count the number of codewords Y at distance *i* from a fixed codeword X.
- This explains the appearance of A(n, d, w)'s in the improvements of LP bound.

### Main result: Improved Schrijver SDP bound

### In Schrijver SDP bound

- We deal with the "triple distance distribution"  $\{x_{i,j}^t\}$ .
- When counting triples  $(X, Y, Z) \in C^3$ , we can first fix two codewords X, Y and then count Z.
- This means we count the number of codewords Z at fixed distances from X and Y.
- Hence,  $A(n, \Lambda, w)$  would be involved, where  $\Lambda$  has two elements.
- Therefore,  $T(w_1, n_1, w_2, n_2, d)$  would appear.

Main result: Improved Schrijver SDP bound-Cont.

**Main Theorem:** For each  $i, j, t \in \{0, ..., n\}$  with  $\binom{n}{i-t, j-t, t} \neq 0$ ,

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

Quick reason:

- We wish to count  $\lambda_{i,j}^t = \left| \begin{cases} (X, Y, Z) \in \mathcal{C}^3 \\ d(Y, Z) = i + j - 2t. \end{cases} \right|_{i,j}^t d(X, Z) = j,$
- Double counting!! We first pick  $(X, Y) \in \mathcal{C}^2$  such that d(X, Y) = i. Then the number of Z such that  $(X, Y, Z) \in \left\{ \left. (X, Y, Z) \in \mathcal{C}^3 \right| \begin{array}{l} d(X, Y) = i, d(X, Z) = j, \\ d(Y, Z) = i + j - 2t. \end{array} \right\}$  is

less that or equal to  $A(n, \Lambda, d)$  where  $\Lambda = \{X, j\}, (Y, i + j - 2t)\}.$  And this value is T(t, i, j - t, n - i, d).

• Now summing over all pairs  $(X, Y) \in \mathcal{C}^2$  such that d(X, Y) = i.

**Corollary:** For each  $j = 0, \ldots, n$ ,

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

Main result: Improved Schrijver SDP bound-Cont.

Recall the conditions of Schrijver SDP bound

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

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

#### **Remark:**

- The Theorem improves the condition  $x_{i,j}^t \leq x_{i,0}^0$  since  $\frac{T(t,i,j-t,n-i,d)}{\binom{i}{t}\binom{n-i}{j-t}}$  is much less than 1 in general.
- The Corollary says that  $x_{i,0}^0 + x_{j,0}^0 \le 1 + x_{i,j}^t$  for all i, j.

#### Main result: Improved Schrijver SDP bound-Cont.

#### More linear constraints

- Since  $B_i = \binom{n}{i} x_{i,0}^0$  for all *i*, all linear constraints on  $B_i$ 's (improvements of LP bound) can be used in SDP bound.
- More linear constraints on  $x_{i,j}^t$ 's have been (and are being) studied (but less hope to make more improvements).

### New upper bounds on A(n,d)

#### Result Improved upper bounds on A(n, d)known known improved new Schijver Schrijver lower upper upper bound bound bound bound bound d n

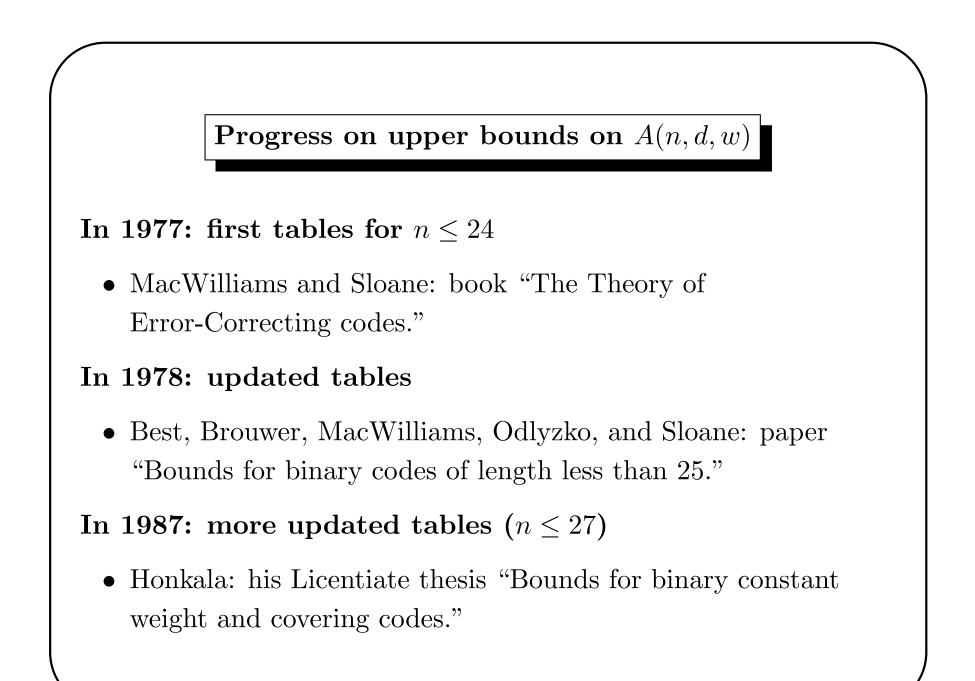
## Remark

#### Final remark:

- We in fact get seven new upper bounds on A(n,d) (for  $n \leq 28$ ).
- However, five of them have been improved by D.C. Gijswijt, H.D. Mittelmann, A. Schrijver, "Semidefinite code bounds based on quadruple distances".
- Two new upper bounds are:

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

#### III. Improvements on optimal constant-weight codes



**Progress on bounds on** A(n, d, w)

In 1990: update tables for  $n \leq 28$ 

• Brouwer, Shearer, Sloane, and Smith: paper "A new table of constant weight codes."

#### In 2000: Improved upper bounds

• Agrell, Vardy, and Zeger: paper "Upper bounds for constant-weight codes."

#### In 2005: Using semidefinite programming

• Schrijver: paper "New code upper bounds from the Terwilliger algebra and semidefinite programming."

#### Our improvements

Main results We give two kind of improvements:

- One intersects the improvement of the Delsarte's linear programming bound in the paper "Upper bounds for constant-weight codes" (in 2000.)
- The other is an improvement of the Schrijver's semidefinite programming bound in the paper "New code upper bounds from the Terwilliger algebra and semidefinite programming" (in 2005.)

#### The first improvement

#### The key point

- In 2000, Agrell, Vardy, and Zeger showed that  $B_i$  and  $B_j$  (for suitable *i* and *j*) have a linear "relation". This gives a linear constraint on  $B_i$  and  $B_j$  which improves the Delsarte's LP bound.
- We show that  $B_i$ 's for  $i \in H$  (with  $|H| \ge 2$ ) also have a linear "relation". For  $n \le 28$ , with |H| = 3, new upper bounds on A(n, d, w) are obtained.
- The improvement comes from the observation that: the existence of a codeword at distance *i* from a fixed codeword *X* will "effect" not only the number of codewords at distance *j* from *X* (showed by Agrell, Vardy, and Zeger) but also the number of codewords at distance *k* from *X*, etc..

#### The first improvement-Cont.

#### Example

- Consider A(27, 8, 13).
- By the result of Agrell, Vardy, and Zeger,

 $B_{22} + 6B_{24} \le 26$ ,  $B_{22} + 26B_{26} \le 26$ ,  $B_{24} + B_{26} \le 1$ .

• Our result gives

$$B_{22} + 6B_{24} + 26B_{26} \le 26, \quad B_{24} + B_{26} \le 1.$$

• We get  $A(27, 8, 13) \leq 11904$ . This improves the upper bound of Agrell, Vardy, and Zeger:  $A(27, 8, 13) \leq 11991$ , and the best upper bound of Schrijver:  $A(27, 8, 13) \leq 11981$ .

The second improvement

#### Schrijver's semidefinite programming (SDP) bound

• Schrijver's (SDP) bound is based on the "triple distance distribution" of constant-weight codes.

The second improvement

"Triple distance distribution" of constant-weight codes

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

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

where  $\mu_{i,j}^{t,s}$  is the number of triples  $(X, Y, Z) \in \mathcal{C}^3$  with d(X,Y) = 2i, d(X,Z) = 2j, d(Y,Z) = 2(i+j-t-s), and d(X+Y,Z) = w + 2t - 2s.

• Set 
$$y_{i,j}^{t,s} = 0$$
 if either  $\binom{w}{i-t,j-t,t} = 0$  or  $\binom{v}{i-s,j-s,s} = 0$ .

#### The second improvement-Cont.

#### Our improvement

- Schrijver showed that  $y_{i,j}^{t,s} \leq y_{i,0}^{0,0}$  and  $y_{i,0}^{0,0} + y_{j,0}^{0,0} \leq 1 + y_{i,j}^{t,s}$  for every t, s, i, j.
- We improve these by showing that

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

for every t, s, i, j, where  $T(w_1, n_1, w_2, n_2, w_3, n_3, w_4, n_4, d)$  can be defined similarly as A(n, d) (the difference is that each codeword must have the form

$$X = (X_1, X_2, X_3, X_4) \in \mathcal{F}^{n_1} \times \mathcal{F}^{n_2} \times \mathcal{F}^{n_3} \times \mathcal{F}^{n_4}$$
  
with  $wt(X_i) = w_i$  for  $i = 1, \dots, 4$ .

#### The second improvement-Cont.

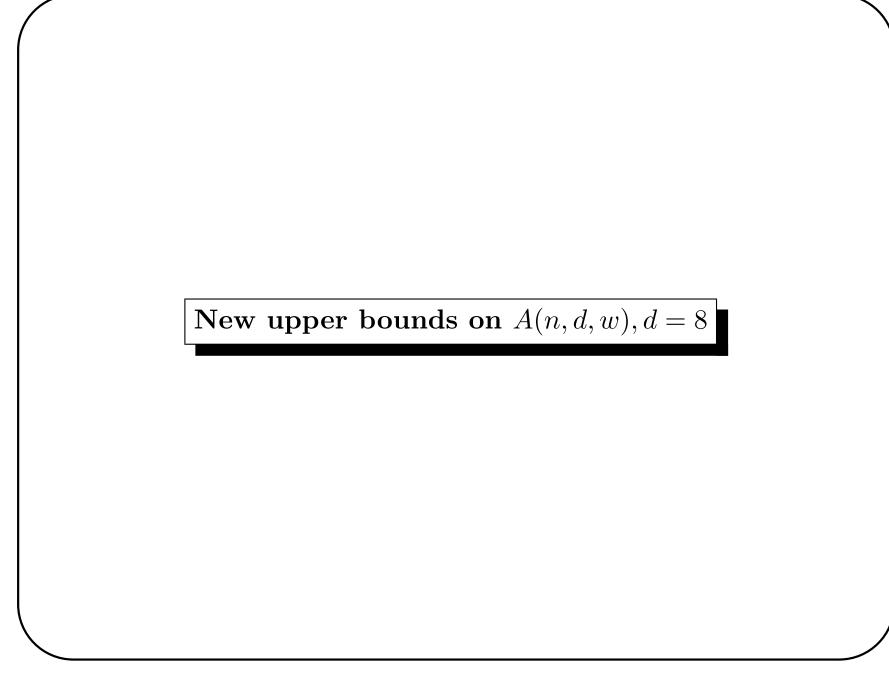
#### The key point

• When counting  $(X, Y, Z) \in C^3$ , we fix (X, Y) first and then count Z. Then we can see that the improvement (naturally) comes from the definition of  $A(n, \Lambda, d)$  for a "special" case of  $|\Lambda| = 4$ . New upper bounds on A(n, d, w), d = 6

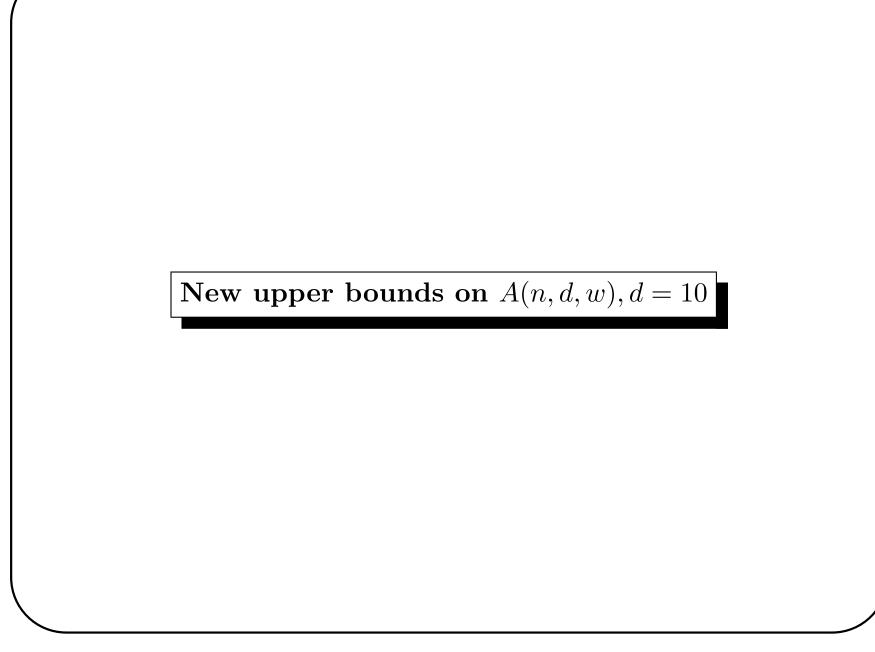
Tables of new upper bounds on A(n, d, w) For  $n \leq 28$ , there are 21 new upper bounds on A(n, d, w) which are listed below.

|    |   |   | lower | upper | new upper | Schrijver |
|----|---|---|-------|-------|-----------|-----------|
| n  | d | W | bound | bound | bound     | bound     |
| 20 | 6 | 8 | 588   | 1107  | 1106      | 1136      |

Table 1: New upper bounds for A(n, d, w)



|    |   |    | lower | upper | new upper | Schrijver |
|----|---|----|-------|-------|-----------|-----------|
| n  | d | W  | bound | bound | bound     | bound     |
| 22 | 8 | 10 | 616   | 634   | 630       | 634       |
| 23 | 8 | 9  | 400   | 707   | 703       | 707       |
| 26 | 8 | 11 | 1988  | 5225  | 5208      | 5225      |
| 27 | 8 | 9  | 1023  | 2914  | 2911      | 2918      |
| 27 | 8 | 11 | 2404  | 7833  | 7754      | 7833      |
| 27 | 8 | 12 | 3335  | 10547 | 10472     | 10697     |
| 27 | 8 | 13 | 4094  | 11981 | 11904     | 11981     |
| 28 | 8 | 11 | 3773  | 11939 | 11896     | 12025     |
| 28 | 8 | 12 | 4927  | 17011 | 17010     | 17011     |
| 28 | 8 | 13 | 6848  | 21152 | 21148     | 21152     |



| [  |    |    |       |       |           |           |
|----|----|----|-------|-------|-----------|-----------|
|    |    |    | lower | upper | new upper | Schrijver |
| n  | d  | W  | bound | bound | bound     | bound     |
| 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  | 1191      | 1201      |
| 27 | 10 | 13 | 405   | 1419  | 1406      | 1419      |
| 28 | 10 | 11 | 308   | 1356  | 1351      | 1356      |

Table 2: New upper bounds for A(n, d, w), d = 12

|    |    |    | lower | upper | new upper | Schrijver |
|----|----|----|-------|-------|-----------|-----------|
| n  | d  | W  | bound | bound | bound     | bound     |
| 25 | 12 | 10 | 28    | 37    | 36        | 37        |

# Thank you!