# On the impossibility of graph secret sharing

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#### Abstract

A perfect secret sharing scheme based on a graph G is a randomized distribution of a secret among the vertices of the graph so that the secret can be recovered from the information assigned to vertices at the endpoints of any edge, while the total information assigned to an independent set of vertices is independent (in statistical sense) of the secret itself.

The efficiency of a scheme is measured by the amount of information the most heavily loaded vertex receives divided by the amount of information in the secret itself. The (worst case) *information ratio* of G is the infimum of this number. We calculate the best lower bound on the information ratio for an infinite family of graphs the celebrated entropy method can give.

We argue that all existing constructions for secret sharing schemes are special cases of the generalized vector space construction. We give direct constructions of this type for the first two members of the family, and show that for the other members no construction exists which would match the bound yielded by the entropy method.

# 1 Introduction

Secret sharing has been investigated in several papers [1, 2, 5, 7, 14, 16, 17, 18, 22] as well as schemes based on graphs [4, 6, 8, 9, 12, 13] just to mention a few. Subsets of the participants are split into *qualified* and *unqualified* ones. A qualified subset can recover the secret, while the total information an unqualified subset has should be (statistically) independent of it. When the scheme is based on a graph, then the participants are the vertices of the graph, and a collection of vertices is qualified if it contains an edge.

The most important property of a scheme is its *efficiency*, namely how many bits the most heavily loaded participant must remember for each bit in the secret. The (worst case) *information* ratio of a graph G is the infimum of the efficiency of all schemes based on G. In the literature the inverse of this number is used and called the *information* rate of G in resemblance to the coding efficiency on noisy channels.

Determining the information ratio for a simple graph could be a very difficult problem cf. [9, 12, 13]. Nevertheless, the ratio was determined exactly for several infinite families of graphs in the above references. Interestingly, all these ratios are of the form 2 - 1/k or k/2 for some positive integer k, and it is an open problem to find a graph with ratio different from these values. In this paper we investigate another infinity family of graphs. We establish the best lower bound the entropy method can give, and show that present-day techniques cannot reach this bound. We formulate an open problem and some conjectures as well.

#### **1.1** Basic notions

In the paper we use the standard techniques and notions, see [4, 6, 8]. For the sake of the reader we briefly repeat some of the definitions.

Let G be a graph. A secret sharing scheme on G is a collection of random variables  $\xi_v$  for all vertices v in G, plus the special random variable  $\xi_s$ . The value of this latter one is the secret, that of the others are the shares. The random variables form a joint distribution. The dealer draws

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from this distribution, and sends the value  $\xi_v$  to the *participant*  $v \in G$ , while keeps the value of  $\xi_s$  secret.

The secret sharing scheme is *perfect* if whenever v, w is an edge in G then  $\xi_v$  and  $\xi_w$  together determine the secret uniquely; while if A is an independent set of vertices (i.e. no edge exist between vertices in A), then the collection  $\{\xi_v : v \in A\}$  is statistically independent from  $\xi_s$ , i.e. the totality of shares this collection has provides zero information about the secret.

Using the usual (Shannon) entropy [11], A determines B iff the entropy of A and AB are the same, while A and B are statistically independent iff the entropy of AB is the sum of the entropies of A and B. (Here and in the sequel we write AB in place of  $A \cup B$ .) The Shannon entropy of A, denoted as  $\mathbf{H}(A)$ , is roughly the number of independent bits necessary to encode the value of A. Applying this notion to the secret sharing we see that the *size of the share* assigned to the participant  $v \in G$  is the entropy of  $\xi_v$ , and the size of the secret is  $\mathbf{H}(\xi_s)$ . Thus the *information ratio* of the graph G is

$$R(G) = \inf\left\{\frac{\max_{v \in G} \mathbf{H}(\xi_v)}{\mathbf{H}(\xi_s)}\right\},\tag{1}$$

where the infimum is taken over all perfect schemes on G.

### 1.2 Proving a lower bound

Let the distribution  $\{\xi_s, \xi_v\}$  be any perfect secret sharing scheme on G. Consider the real-valued function f which assigns the value

$$f(A) = \frac{\mathbf{H}(\{\xi_v : v \in A\})}{\mathbf{H}(\xi_s)}$$
(2)

to the subset A of the vertices. Using standard properties of the entropy function [11, 20, 24], this f has certain properties, namely

 $\begin{array}{ll} (a) & f(A) \geq 0, \ f(\emptyset) = 0 & \mbox{positivity,} \\ (b) & f(B) \geq f(A) \ \mbox{when } B \supseteq A & \mbox{monotonicity,} \\ (c) & f(A) + f(B) \geq f(A \cup B) + f(A \cap B) & \mbox{submodularity.} \end{array}$ 

Furthermore,  $\mathbf{H}(\xi_s \cup \{\xi_v : v \in A\})$  is equal to  $\mathbf{H}(\{\xi_v : v \in A\})$  if A determines the secret, i.e. when A contains an edge; and is equal to the sum of  $\mathbf{H}(\xi_s)$  and  $\mathbf{H}(\{\xi_v : v \in A\})$  when A is independent. This observation leads us to

(d)  $f(B) \ge f(A) + 1$ when  $B \subseteq A$ , A is independent and B is not (e)  $f(A) + f(B) \ge f(A \cup B) + f(A \cap B) + 1$ when  $A \cap B$  are independent but A, B are not.

Properties (d) and (e) are called *strict monotonicity* and *strict submodularity*, respectively.

Unfortunately properties (a)–(e) do not characterize the real-valued functions which can be got from perfect secret sharing schemes. They satisfy further *non Shannon-type* inequalities [24] as well, and no complete characterization is known up to now [20]. Even for the case of four (dependent) random variables we do not have a complete description of the 15 dimensional cone spanned by the entropies of the different subsets of the variables [20].

Now we can describe the celebrated *entropy method* [4, 6, 8]. Take any real-valued function f which satisfies properties (a)–(e) above (and maybe further inequalities which are always true for the entropy function). Suppose we show for some real number r that for all of such functions f,  $\max_{v \in G} f(v) \ge r$ . Then the *information ratio* R(G) of G defined in (1) is at least r.

Observe that properties (a)–(e) are linear inequalities (and, in fact, all further known properties of the entropy are), thus determining the best lower bound is, in fact, an LP problem [10]. The only problem is that both the number of variables and the number of constraints grow exponentially with the size of the graph, which makes the LP problem unsolvable even for relatively small

graphs. We remark for the interested reader that the system of conditions given by (a)-(e) is overdetermined. Even after reducing the conditions [20], the system remains ill-posed which makes further complications.

### 1.3 Proving an upper bound

Typically upper bounds come easily: one has to find an appropriate scheme which realizes the given bound. There are constructions based on some algebraic structure (mainly vector spaces over finite fields) [1, 13, 12, 21], or on geometry (finite projective geometry) [3]. The celebrated, and incredibly effective construction of Stintson [22] can be used to build a scheme from other smaller schemes. van Dijk and al [13] used a slightly different method where the intermediate schemes are not necessarily perfect. Nevertheless, all presently known constructions [1, 5, 17] (even the ones arising from van Dijk's construction or from span programs) are special cases of the following general one.

Let  $\mathbb{F}$  be a vector space (sometimes a weaker structure, such as a module suffices), and assign (non-trivial) linear subspaces of  $\mathbb{F}$  both to the participants and the secret: let  $L_v$  be the subspace assigned to  $v \in G$  and  $L_s$  be the subspace assigned to the secret. These subspaces should have the following property: if vw is an edge in G, then the linear span of  $L_v$  and  $L_w$  should contain (as a subspace)  $L_s$ . If, on the other hand,  $\{v_1, \ldots, v_k\}$  is an independent set (this is always the case when k = 1), then the intersection of the linear span of  $\{L_{v_1}, \ldots, L_{v_k}\}$  and  $L_s$  must be trivial, i.e. the single element subspace  $\{0\}$ .

The dealer chooses an element from  $\mathbb{F}$  uniformly (here we must assume that  $\mathbb{F}$  is finite). The *secret*, i.e. the value of  $\xi_s$  is the orthogonal projection of this random element on  $L_s$ . The *share* of participant  $v \in G$  is the orthogonal projection of the dealer's element on  $L_v$ .

Now, if vw is an edge, then using elementary linear algebra, the secret can be expressed as an appropriate linear combination of the shares. On the other hand, if  $\{v_1, \ldots, v_k\}$  is an independent subset of vertices, then the linear span of  $\{L_{v_1}, \ldots, L_{v_k}\}$  and the subspace  $L_s$  intersect in the zero vector, thus projection on the first one gives no information at all on the value of projection on the other. (This is the second point where the finiteness of  $\mathbb{F}$  plays a crucial role.)

The amount of information (i.e. entropy) in the secret is proportional to the dimension of  $L_s$ , and the information v gets is proportional to  $\dim(L_v)$ . Thus the *ratio* of this construction is

$$\frac{\max_{v\in G}\dim(L_v)}{\dim(L_s)}$$

The total randomness the dealer needs to produce the shares is proportional to the dimension of the whole vector space  $\mathbb{F}$ .

Looking at this construction more carefully, the function f defined in (2) takes the same value as the ratio of the dimensions of the corresponding subspaces:

$$f(A) = \frac{\mathbf{H}(\{\xi_v : v \in A\})}{\mathbf{H}(\xi_s)} = \frac{\dim(\langle L_v : v \in A \rangle)}{\dim(L_s)}$$

Linear subspaces of a vector space form a matroid [23]. However not all matroids can be represented this way. Matroids arising from linear subspaces satisfy the so-called *Ingleton inequality* [15], which not all matroids, and not all functions arising from entropy, do:

(f) 
$$f(AC) + f(AD) + f(BC) + f(BD) + f(CD) \ge$$
  
 $f(C) + f(D) + f(ACD) + f(BCD) + f(AB)$ . (3)

In particular, this inequality is not a consequence of the inequalities (a)–(e) discussed above, but it always holds for all existing secret sharing constructions.

# 2 The graph family

One of the smallest graphs where no exact information ratio was known for a long time [12] is the following. It has six vertices,  $v_1, v_2, v_3$ , and  $w_1, w_2, w_3$ . The first three vertices form a triangle, furthermore only  $v_i$  and  $w_i$  are connected.

Using the entropy method sketched in section 1.2, an LP package was used to get the optimal bound for this graph, which turned out to be 7/4. There is an easy construction with ratio 2 using Stinson's decomposition method [22], but the exact value was not known for some time. The first published construction with ratio 7/4 can be found in [13]. This graph is clearly an element of the following infinite family of graphs:

Let  $G_n$  have vertices  $v_1, \ldots, v_n$  and  $w_1, \ldots, w_n$ . The edges are  $v_i v_j$  for each pair *i* and *j*, furthermore only  $v_i$  and  $w_i$  are connected. That is,  $G_n$  is a complete graph on *n* vertices and each vertex is connected to an extra vertex from an independent set of size *n*. The above graph is  $G_3$ , while  $G_2$  is the path of length 4 (a complete graph on 2 vertices, plus two additional vertices). The information ratio of  $G_2$  is 3/2 [4], while that of  $G_3$  is 7/4. Using an LP package we found that the entropy method yields the lower bound 15/8 for  $G_4$ . This data supported the conjecture that the ratio of  $G_n$  is at least  $(2^n - 1)/2^{n-1} = 2 - 1/2^{n-1}$ . In section 3 we show that indeed this is the case, and, furthermore, this is the best value what the entropy method can give.

In section 4 we give a novel construction for  $G_3$  which matches the lower bound, finally in section 5 we show that there exists no similar construction which would work for  $G_4$ , and, consequently, for other graphs in this family. In the last section we discuss the intuition that no vector-space construction can exist in general for this graph family. We also list some open problems.

# **3** Lower bound for $G_n$

Let  $G_n$  be the graph defined above. Among its 2n vertices  $v_1, \ldots, v_n$  form a complete graph, while the vertex  $w_i$  is connected to  $v_i$  only. The set of vertices  $\{v_1, \ldots, v_n\}$  is denoted by V, while set of the other is denoted by W, where W is an independent set (i.e. it contains no edges).

As explained in section 1.2, let f be a real function assigning non-negative values to subsets of vertices so that f satisfies properties (a)–(e) listed there. Our goal is to give the best possible lower estimate for  $\max_{v \in V \cup W} f(v)$ . We start with a lemma. As customary, we leave out the {} and  $\cup$  signs, and write, e.g., vX for the set {v}  $\cup X$ .

**Lemma 1** Let X be a subset of W,  $w \in W - X$ ,  $a, b \in V$  so that a is not connected to any vertex in  $X \cup \{w\}$ , while b is connected to w. Then

$$f(aX) - f(X) + f(bX) - f(X) \ge f(awX) - f(wX) + 2$$

**Proof** Observe that awX is independent, while abwX is not. Thus property (d) gives

$$f(abwX) \ge f(awX) + 1.$$

As bX is independent, abX and bwX are not, the strict submodularity property (e) gives the first line below. Other lines are instances of the submodularity property (b):

$$\begin{array}{rcl} f(abX) + f(bwX) & \geq & f(bX) + f(abwX) + 1 \\ f(aX) + f(bX) & \geq & f(abX) + f(X) \\ f(wX) + f(bX) & \geq & f(X) + f(bwX) \end{array}$$

Adding up these inequalities we get the claim of the lemma.

Using the lemma with X as the empty set we get

$$f(v_2) + f(v_1) \ge f(v_2w_1) - f(w_1) + 2,$$

and similarly

$$f(v_3) + f(v_1) \ge f(v_3w_1) - f(w_1) + 2.$$

Adding these up and using the lemma again with  $X = \{w_1\}$  we have

$$f(v_3) + f(v_2) + 2f(v_1) \ge f(v_3w_2w_1) - f(w_2w_1) + 2 \cdot 2 + 2.$$

Similar reasoning gives

$$f(v_4) + f(v_2) + 2f(v_1) \ge f(v_4w_2w_1) - f(w_2w_1) + 2 \cdot 2 + 2,$$

or, one can argue, the conditions are invariant under swapping  $v_3$  and  $v_4$  and  $w_3$  and  $w_4$  (and keeping all other vertices fixed), thus all results are also invariant for this variable change. Applying the lemma again we arrive at

$$f(v_4) + f(v_3) + 2f(v_2) + 4f(v_1) \ge f(v_4w_3w_2w_1) - f(w_3w_2w_1) + 2 \cdot 2^2 + 2 \cdot 2 + 2.$$

Here we can replace  $v_4$  by  $v_5$ , and continue the same way until there are no more vertices in V:

$$f(v_n) + f(v_{n-1}) + 2f(v_{n-2}) + 2^2 f(v_{n-3}) + \dots + 2^{n-3} f(v_2) + 2^{n-2} f(v_1) \ge$$
  
 
$$\ge f(v_n w_{n-1} \dots w_2 w_1) - f(w_{n-1} \dots w_2 w_1) + 2(2^{n-1} - 1).$$

Let  $Y = \{w_{n-1}, \dots, w_2, w_1\}$ , then

$$f(v_nY) - f(Y) \ge f(v_nw_nY) - f(w_nY) \ge 1.$$

Here the first inequality is an equivalent form of submodularity (c), while the second one is the strict monotonicity property (d). Consequently

$$f(v_n) + f(v_{n-1}) + 2f(v_{n-2}) + \ldots + 2^{n-2}f(v_1) \ge 2^n - 1.$$

By symmetry the same inequality is valid for all circular shifts of the vertices. There are n such instances all together, adding them up each  $f(v_i)$  will have coefficient

$$1 + 1 + 2 + 4 + \ldots + 2^{n-2} = 2^{n-1},$$

consequently the sum is

$$2^{n-1}(f(v_1) + f(v_2) + \ldots + f(v_n)) \ge n(2^n - 1).$$
(4)

Therefore not all of the values  $f(v_i)$  can be smaller than  $(2^n - 1)/2^{n-1} = 2 - 2^{-n+1}$ . That is, we have proved the following

**Theorem 2** The ratio of the graph  $G_n$  is at least  $2 - 2^{-n+1}$ .

In section 5 we shall need the following result which can be proved analogously.

**Lemma 3** Suppose  $n \ge 4$ . For some  $3 \le k \le n$  we have

$$f(v_k w_2 w_1) - f(w_2 w_1) \ge 2 - 2^{-n+3}.$$

**Proof** Let us denote the value  $f(v_k w_2 w_1) - f(w_2 w_1)$  by  $a_k$ . As in the previous proof, Lemma 1 gives

$$a_4 + a_3 \ge f(v_4 w_3 w_2 w_1) - f(w_3 w_2 w_1) + 2,$$

and also

$$a_5 + a_3 \ge f(v_5 w_3 w_2 w_1) - f(w_3 w_2 w_1) + 2.$$

Adding these up and applying the lemma again we get

$$a_5 + a_4 + 2a_3 \ge f(v_5w_4w_3w_2w_1) - f(w_4w_3w_2w_1) + 2 \cdot 2 + 2.$$

Continuing as above, we get

$$a_n + a_{n-1} + 2a_{n-2} + \ldots + 2^{n-5}a_4 + 2^{n-4}a_3 \ge \ge f(v_n w_{n-1} \dots w_2 w_1) - f(w_{n-1} \dots w_2 w_1) + 2^{n-2} - 2 \ge \ge 1 + 2^{n-2} - 2 = 2^{n-2} - 1.$$

Making a cyclic shift of the vertices  $v_k, \ldots, v_3$ , we get

$$a_3 + a_n + 2a_{n-1} + \ldots + 2^{n-5}a_5 + 2^{n-4}a_4 \ge 2^{n-2} - 1.$$

Adding up all of these n-2 inequalities,

$$2^{n-3}(a_n + a_{n-1} + \ldots + a_3) \ge (n-2)(2^{n-2} - 1)$$

from where the claim of the lemma follows.

The lower bound in Theorem 2 is the best possible one what the entropy method sketched in section 1.2 can give. To show it we present a function f with properties (a)–(e) which, in addition, satisfies  $f(v) \leq 2 - 2^{-n+1}$  for all vertices v in the graph. In fact, we'll have equality for vertices in V, while f(w) = 1 for vertices of degree one.

This function f should be defined for all subsets of the vertices. Let the set of vertices of  $G_n$  be  $V \cup W$ . With each  $A \subseteq V \cup W$  we associate three non-negative integers  $i_A, j_A$  and  $k_A$  as follows. A contains exactly  $j_A$  pairs  $v_i w_i$  where  $v_i$  and  $w_i$  are connected,  $v_i \in V$  and  $w_i \in W$ . Apart from these vertices there are  $i_A$  vertices of A in V, and  $k_A$  vertices of A in W.

Now  $|A| = i_A + 2j_A + k_A$ , and A is independent iff  $i_A \leq 1$  and  $j_A = 0$ . Let furthermore  $\ell_A = i_A + j_A + k_A$ , obviously  $\ell_A \leq n$ . Define the function f on all subsets of the vertices as follows:

$$f(A) = \begin{cases} \ell_A & \text{if } i_A + j_A = 0, \\ \ell_A + 1 - 2^{-n + \ell_A} & \text{if } i_A + j_A > 0 \text{ and } A \text{ is independent}, \\ \ell_A + 2 - 2^{-n + \ell_A} & \text{otherwise.} \end{cases}$$
(5)

It is a tedious but otherwise trivial task to check that indeed this f satisfies properties (a)–(e) for all subsets of the vertices. When  $A = \{v\}$  and v is a V, then  $i_A = 1$ ,  $j_A = k_A = 0$ , thus  $\ell_A = 1$ and  $f(v) = 1 + 1 - 2^{-n+1}$ . Similarly, if  $A = \{w\}$  with  $w \in W$  then  $i_A = j_A = 0$ ,  $k_A = 1$ , thus f(w) = 1 (first case of the definition in (5)).

### 4 A novel construction

In this section we give a construction which matches the corresponding lower bound for the graphs  $G_2$  and  $G_3$ , and show how to generalize it for arbitrary n to yield the upper bound 2.

Our construction follows the idea outlined in section 1.3. Namely, we start with a highdimensional vector space  $\mathbb{F}$ , and assign linear subspaces to the vertices *and* the secret so that

- if v and w are connected, then the linear span of the subspaces  $L_v$  and  $L_w$  contain the subspace  $L_s$  assigned to the secret, and
- whenever  $\{v_1, \ldots, v_k\}$  is an independent set then the linear span of  $\{L_{v_1}, \ldots, L_{v_k}\}$  intersects  $L_s$  in the null space  $\{0\}$ .

Having such subspaces, we can construct a perfect secret sharing scheme with ratio

$$\frac{\max_{v\in G}\dim(L_v)}{\dim(L_s)}.$$

In our case the graph  $G_n$  has vertices  $v_i$  and  $w_i$  for  $1 \le i \le n$ , where  $V = \{v_1, \ldots, v_n\}$  is a complete graph, while  $\{w_1, \ldots, w_n\}$  is empty (i.e. independent).  $\mathbb{F}$  will have dimension d(n+1), and all subspaces will be given as the linear span of certain vectors.

Each element in  $\mathbb{F}$  is a vector with d(n+1) coordinates. We split these coordinates into n+1 groups of coordinates d each. We define k vectors from  $\mathbb{F}$  as a sequence of n+1 matrices each of size  $k \times d$ . As usually,  $I = I_d$  is the unit  $d \times d$  matrix: it has 1 in the diagonal elements and zero elsewhere.

The secret is assigned the subspace spanned by the d vectors of the form  $I, \ldots, I, I$ :

$$L_s = (I, \ldots, I, I)$$

where we have exactly n + 1 unit matrices here. As these vectors are linearly independent, the dimension of  $L_s$  is d.

Vertices in the independent set  $\{w_1, \ldots, w_n\}$  will be assigned a subspace generated by the d vectors

$$L_{w_i} = (0, \ldots, 0, I, 0, \ldots, 0, 0)$$

where the only I block is at the *i*-th position. Here  $\dim(L_{w_i}) = d$  again, and the linear span of all subspaces  $L_{w_i}$  contain those vectors where all coordinates in the last, (n+1)-st block are zero. As any non-trivial linear combination of  $L_s$  has non-zero coordinate in each block, consequently

$$\langle L_{w_1}, \ldots, L_{w_n} \rangle \cap L_s = \{0\}$$

thus satisfying the second requirement for the independent set W.

Next we assign linear spaces to the remaining vertices  $v_i$ . These subspaces should satisfy the following requirements:

- 1. the span of  $L_{v_i}$  and  $L_{w_i}$  must contain  $L_s$ ,
- 2. the span of  $L_{v_i}$  with  $\{L_{w_j} : j \neq i\}$  should avoid  $L_s$ , finally
- 3. the span of two different  $L_{v_i}$  and  $L_{v_j}$  should contain  $L_s$  again.

To satisfy the first condition we include in  $L_{v_i}$  the vectors

$$(I,\ldots,I,0,I,\ldots,I,I)$$

where only the *i*-th block is zero. The sum of the *j*-th vector from  $L_{w_i}$  and the *j*-th vector from  $L_{v_i}$  gives the generating elements of  $L_s$ , i.e. the linear span of  $L_{v_i}$  and  $L_{v_i}$  contains  $L_s$  as required.

To satisfy the second condition, we stipulate that all vectors in  $L_{v_i}$  should have zero coordinate in the *i*-th block. Then the linear span of  $L_{v_i}$  with all other  $L_{w_j}$ 's with  $j \neq i$  has zero coordinate in this block, consequently contains only the all zero element from  $L_s$ .

The difficulty comes with the third condition. First, we show how to add d further vectors to each  $L_{v_i}$  to satisfy it. Then we show how to reduce the number of added vectors when n = 2 or n = 3.

Let  $M_1, M_2, \ldots, M_n$  be  $d \times d$  matrices so that  $M_i - M_j$  has full rank whenever  $i \neq j$ . This is the case, for example, when we choose  $M_i = \lambda_i I$  for different constants  $\lambda_i$ . We add to the generating set of  $L_{v_i}$  the vectors

$$(M_i - M_0, M_i - M_1, \dots, M_i - M_{i-1}, 0, M_i - M_{i+1}, \dots, M_i - M_n, M_i).$$

As the *i*-th block is all zero, the second condition holds. To check that the third condition holds as well, observe that the difference of the latter d vectors assigned to  $v_i$  and  $v_j$  is

$$(M_i - M_j, M_i - M_j, \dots, M_i - M_j, M_i - M_j),$$

and since  $M_i - M_j$  has full rank, the linear span of these vectors contain the generating vectors of  $L_s$  as well.

In this construction each  $L_{v_i}$  is generated by 2*d* linearly independent vectors, thus dim $(L_{v_i}) = 2d$ , while dim $(L_s) = d$ , which shows that it has ratio 2.

To reduce the dimension of  $L_{v_i}$  we look at the first d generating vectors more carefully:

 $(I,\ldots,I,0,I,\ldots,I,I)$ .

The linear span of  $L_{v_i}$  and  $L_{v_i}$  must contain all vectors in the generating set of  $L_s$ , i.e. the vectors

$$(I,\ldots,I,I,I,\ldots,I,I),$$

which happens iff it contains the vectors

$$(0,\ldots,0,I,0,\ldots,0,0)$$

where the only I occurs at the *i*-th position. Now the linear span of  $L_{v_i}$  and  $L_{v_i}$  definitely contains

$$(0,\ldots,0,I,0,\ldots,0,-I,0,\ldots,0,0)$$

where I is at the *i*-th block, and -I is at the *j*-th block. Then it also contains the vectors of the form

$$(0,\ldots,0,\mathbf{x},0,\ldots,0,-\mathbf{x},0,\ldots,0,0)$$

for an arbitrary *d*-dimensional vector  $\mathbf{x}$ , which means that in the linear span we can move the content of the *i*-th block into the *j*-th block, effectively zeroing all elements in one of the block. We shall use this observation to reduce the dimension of  $L_{v_i}$ .

#### 4.1 The case of n = 2

When n = 2 we will choose d = 2 and the linearly independent 2-dimensional vectors **x** and **y**. The vectors which span the subspaces  $L_{v_1}$  and  $L_{v_2}$ , respectively, are

(	0, 0,	0, 1,	0,1	)	(	0, 1,	0, 0,	0, 1	)
(	0, 0,	1, 0,	1,0	)	(	1, 0,	0, 0,	1,0	)
(	0, 0,	$\mathbf{x}$ ,	0, 0	)	(	у,	0, 0,	0, 0	)

It is clear that both spaces have dimension 3, moreover their linear span contains the vectors (I, -I, 0) and  $(0, \mathbf{x}, 0)$ , thus also the vector  $(\mathbf{x}, 0, 0)$  as explained above. This together the vector  $(\mathbf{y}, 0, 0)$  from  $v_2$ ' set gives all vectors in the linear span of (I, 0, 0), as was required.

### 4.2 The case of n = 3

In this case we choose d = 4 and six 4-dimensional vectors  $\mathbf{x}_1, \ldots, \mathbf{x}_6$  such that any four of them has full rank. The subspaces assigned to  $v_1, v_2$  and  $v_3$  are generated by seven vectors as follows:

(	0	Ι	Ι	Ι	)	(	Ι	0	Ι	Ι	)	(	Ι	Ι	0	Ι	)
(	0	$\mathbf{x}_2$	0	0	)	(	$\mathbf{x}_1$	0	0	0	)	(	$\mathbf{x}_1$	$\mathbf{x}_2$	0	0	)
(	0	0	$\mathbf{x}_4$	0	)	(	$\mathbf{x}_3$	0	$\mathbf{x}_4$	0	)	(	$\mathbf{x}_3$	0	0	0	)
(	0	$\mathbf{x}_5$	$\mathbf{x}_6$	0	)	(	0	0	$\mathbf{x}_6$	0	)	(	0	$\mathbf{x}_5$	0	0	)

This construction has ratio 7/4, and it works indeed. For example, in the linear span of  $L_{v_1} \cup L_{v_2}$  we have the four vectors  $(\mathbf{x}_1, 0, 0, 0)$ ,  $(0, \mathbf{x}_2, 0, 0)$ ,  $(\mathbf{x}_3, 0, 0, 0)$  and  $(0, \mathbf{x}_5, 0, 0)$ . Indeed the first two are explicitly given, the third and fourth ones can be got as the difference of one-one vectors from the assigned subspaces:

$$\begin{aligned} (\mathbf{x}_3, 0, 0, 0) &= (\mathbf{x}_3, 0, \mathbf{x}_4, 0) - (0, 0, \mathbf{x}_4, 0) \\ (0, \mathbf{x}_5, 0, 0) &= (0, \mathbf{x}_5, \mathbf{x}_6, 0) - (0, 0, \mathbf{x}_6, 0) \end{aligned}$$

As (I, -I, 0, 0) is also in the span, so is  $(\mathbf{x}_2, -\mathbf{x}_2, 0, 0)$  and then  $(\mathbf{x}_2, -\mathbf{x}_2, 0, 0) + (0, \mathbf{x}_2, 0, 0) = (\mathbf{x}_2, 0, 0, 0)$  is there as well. Consequently all vectors  $(\mathbf{x}_1, 0, 0, 0), (\mathbf{x}_2, 0, 0, 0), (\mathbf{x}_3, 0, 0, 0)$  and  $(\mathbf{x}_5, 0, 0, 0)$  are in the linear span of  $L_{v_1} \cup L_{v_2}$ , thus there are all vectors of the form (I, 0, 0, 0), as was required.

### 5 Impossibility of tight vector space construction

The *Ingleton inequality* (f) cited in section 1.3 holds for all representable matroids, and, in general, for all secret sharing schemes based on the general construction outlined in sections 1.3 and 4, see [15]. For the sake of the reader we repeat the inequality here:

(f) 
$$f(AC) + f(AD) + f(BC) + f(BD) + f(CD) \ge$$
  
 $f(C) + f(D) + f(ACD) + f(BCD) + f(AB)$ .

Using this inequality we can check that the bound  $2-2^{-n+1}$  got in Theorem 2 in *not* achievable by a vector space construction for  $n \ge 4$ . Should such a construction exist, the extremal point of the LP problem given in (5) would satisfy the Ingleton inequality (f) as well, which it does not. Apply (f) with the following cast:

$$A = v_2 w_1, \quad B = v_3 w_1, \quad C = v_1 w_1, \quad D = w_1 w_4$$

where, as usual,  $v_i \in V$ ,  $w_i \in W$  and  $v_i$  and  $w_i$  are connected. The left hand side value of (f) is

$$(4 - 2^{-n+2}) + (4 - 2^{-n+3}) + (4 - 2^{-n+2}) + (4 - 2^{-n+3}) + (4 - 2^{-n+2}) =$$
  
= 20 - 14 \cdot 2^{-n+1}.

which is computed from values given in (5), while the value of the right hand side of (f) is

$$(3 - 2^{-n+1}) + 2 + (5 - 2^{-n+3}) + (5 - 2^{-n+3}) + (5 - 2^{-n+3}) = 20 - 13 \cdot 2^{-n+1}.$$

Consequently the value of the left hand side of (f) does *not* exceed that of the right hand side, i.e. the Ingleton inequality does not hold for this case.

Unfortunately we are not done. Equation (5) gives a feasible solution of the LP problem defined by all conditions in (a)–(e), and by the result in section 3 the solution (5) is on the boundary. Showing that this point does not satisfy a particular instance of the Ingleton inequality does not necessarily mean that another extremal solution wouldn't do it. So we prove the following stronger statement:

**Theorem 4** Let  $n \ge 4$  and suppose the perfect secret sharing scheme on  $G_n$  is based on a vector space construction. Then the ratio is at least  $2 - 2^{-n+1} + 0.2 \cdot 2^{-n+1}$ , i.e. exceeds the lower bound of Theorem 2 by  $0.2 \cdot 2^{-n+1}$ .

**Proof** Let f be any real valued function satisfying conditions (a)–(e) and all instances of the Ingleton inequality (f) where the subsets might contain the secret as well. We show that in this case  $f(v) \ge 2 - 0.8 \cdot 2^{-n+1}$  for some vertex v which proves the Theorem.

As in section 3 the vertices of  $G_n$  are denoted by  $v_i$ ,  $w_i$  for  $1 \le i \le n$  so that  $v_i$  and  $w_i$  are connected, the subset  $\{v_1, \ldots, v_n\}$  is a complete graph, while  $\{w_1, \ldots, w_n\}$  is empty.

By Lemma 3 we may assume that

$$f(v_3w_2w_1) - f(w_2w_1) \ge 2 - 2^{-n+3} \tag{6}$$

by relabeling the vertices if necessary. Let moreover  $v_0, w_0$  be the vertices  $v_4, w_4$ , respectively.

Claim 5  $f(v_1v_0) + f(w_1w_0) \ge f(v_1w_1w_0) + 2.$ 

**Proof** The claim follows from the following sequence of inequalities:

$$f(v_{1}v_{\circ}) \stackrel{(1)}{\geq} f(v_{1}v_{\circ}) + (f(v_{\circ}w_{\circ}) - f(v_{\circ}) - f(w_{\circ})) \geq \\ \stackrel{(2)}{\geq} f(v_{1}v_{\circ}) + (f(v_{1}v_{\circ}w_{\circ}) - f(v_{1}v_{\circ}) + 1) - f(w_{\circ}) = \\ = f(v_{1}v_{\circ}w_{\circ}) - f(w_{\circ}) + 1 \geq \\ \stackrel{(3)}{\geq} (f(v_{1}w_{\circ}) + 1) - f(w_{\circ}) + 1 \geq \\ \stackrel{(4)}{\geq} f(v_{1}w_{1}v_{\circ}) - f(w_{1}w_{\circ}) + 2.$$

Here (1) follows from the submodular property  $f(v_{\circ}) + f(w_{\circ}) \ge f(v_{\circ}w_{\circ})$ ; (2) is the strict submodularity as both  $v_1v_{\circ}$  and  $v_{\circ}w_{\circ}$  are edges; (3) is strict monotonicity using that  $v_1v_{\circ}$  is and edge and  $v_1w_{\circ}$  is empty, finally (4) is the submodularity.

Claim 6  $f(v_3v_2w_1) - f(w_1) \ge 4 - 2^{-n+3}$ .

**Proof** Similarly as before, this is a consequence of the following sequence of inequalities:

$$\begin{aligned} f(v_3v_2w_1) - f(w_1) &= \\ &= \left(f(v_3v_2w_1) - f(v_2w_1)\right) + \left(f(v_2w_1) - f(w_1)\right) \ge \\ &\stackrel{(1)}{\ge} \left(f(v_3v_2w_2w_1) - f(v_2w_2w_1) + 1\right) + \left(f(v_2w_2w_1) - f(w_2w_1)\right) = \\ &= f(v_3v_2w_2w_1) - f(w_2w_1) + 1 = \\ &= \left(f(v_3v_2w_2w_1) - f(v_3w_2w_1)\right) + \left(f(v_3w_2w_1) - f(w_2w_1)\right) + 1 \ge \\ &\stackrel{(2)}{\ge} 1 + \left(2 - 2^{-n+3}\right) + 1. \end{aligned}$$

At (1) we applied strict submodularity and submodularity, while (2) follows from the choice of the indices of the vertices (cf. (6)), and from the strict monotonicity.  $\Box$ 

Turning to the proof of the Theorem, we shall use a single instance of the Ingleton inequality (f) for the same case as at the beginning of this section, namely

$$A = v_2 w_1, \quad B = v_3 w_1, \quad C = v_1 w_4, \quad D = w_1 w_4,$$

and then (f) becomes

$$\begin{aligned} f(v_2v_1w_1) + f(v_2w_1w_0) + f(v_3v_1w_1) + f(v_3w_1w_0) + f(v_1w_1w_0) \geq \\ \geq f(v_1w_1) + f(w_1w_0) + f(v_2v_1w_1w_0) + f(v_3v_1w_1w_0) + f(v_3v_2w_1). \end{aligned}$$

We continue with a series of inequalities which will be added to this one:

$$\begin{array}{rcl}
f(v_2v_1w_1w_0) &\stackrel{(1)}{\geq} & f(v_2w_1w_0) + 1 \\
f(v_3v_1w_1w_0) &\stackrel{(1)}{\geq} & f(v_3w_1w_0) + 1 \\
f(v_1w_1) + f(v_2v_1) &\stackrel{(2)}{\geq} & f(v_1) + f(v_2v_1w_1) + 1 \\
f(v_1w_1) + f(v_3v_1) &\stackrel{(2)}{\geq} & f(v_1) + f(v_3v_1w_1) + 1 \\
f(v_1) + f(w_1) &\stackrel{(2)}{\geq} & f(v_1w_1) \\
f(v_1v_0) + f(w_1w_0) &\stackrel{(3)}{\geq} & f(v_1w_1w_0) + 2 \\
f(v_3v_2w_1) &\stackrel{(4)}{\geq} & f(w_1) + 4 - 2^{-n+3}
\end{array}$$

Here (1) is strict monotonicity, (2) is strict submodularity, (3) comes from Claim 5, and (4) comes from Claim 6. The sum is

$$f(v_2v_1) + f(v_3v_1) + f(v_1v_0) \ge f(v_1) + 10 - 2^{-n+3}.$$

From here, using that  $f(v_i v_j) \ge f(v_i) + f(v_j)$ , we arrive at

$$2f(v_1) + f(v_2) + f(v_3) + f(v_0) \ge 10 - 2^{-n+3} = 10 - 4 \cdot 2^{-n+1}.$$

It means that not all of the values  $f(v_1)$ ,  $f(v_2)$ ,  $f(v_3)$ ,  $f(v_0)$  can be below  $2 - 0.8 \cdot 2^{-n+1}$ , proving the Theorem.

### 6 Conclusion

We defined an sequence of graphs  $G_n$ , and considered perfect secret sharing schemes based on them. We established the best lower bound on the efficiency of the schemes the entropy method can give, and matched that lower bound for  $G_2$  and  $G_3$  by a novel construction. We also proved in Theorem 4 that no similar construction exists for other members of the family: any secret sharing scheme for  $G_n$  based on linear construction must have a strictly larger rate than the entropy method gives.

As all presently known constructions are based on some linear coding, they are subject to our result. Without breakthrough new results we cannot hope for a construction, even for  $G_4$ , which would match the lower bound  $2 - 2^{-n+1}$ .

**Problem 7** Show that  $R(G_4) > 2 - 2^{-n+1}$ , i.e. there is no perfect secret sharing scheme on  $G_4$  which would match the entropy bound.

There are further, non-Shannon type inequalities, cf. [24], which the function f in section 1.2 must satisfy beyond (a)–(e) enlisted there. The extremal point found in (5) satisfies these extra inequalities as well, thus they do not help in solving Problem 7 as they did in [1].

In section 4 we showed how to construct a scheme with ratio 2 for arbitrary n. Our result in Theorem 4 indicates, that any vector space construction must have higher ratio than the absolutely minimum given by the entropy method. This ratio, however, is still below 2, but we were unable to construct any such scheme in general. It is not hard to see that the scheme must follow the pattern outlined there, namely subspaces assigned to the 1-degree vertices can be assumed to be pairwise orthogonal, have similar connection to the secret subspace, and subspaces assigned to other vertices must follow similar pattern as well.

Also, if the ratio is below 2 then it is exponentially close to 2, which means that the vector space dimension must also be exponential (in n). But this contradicts to the intuition that we do not need more dimensions than the number of minimal qualified subsets multiplied by the dimension of the secret, which is definitely below  $n^3$ .

**Problem 8** Find a perfect secret sharing scheme on  $G_n$ ,  $n \ge 4$  with ratio strictly below 2, or show that no such a scheme exists.

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