

$(\epsilon = 1, \ell = 1)$
 nodes and 16⁶
 ≤ 42 and $r \leq 1$
 esmanian codes
 While writing
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5. Combinatorial Results

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find better solution

than this space
 (more intro?)

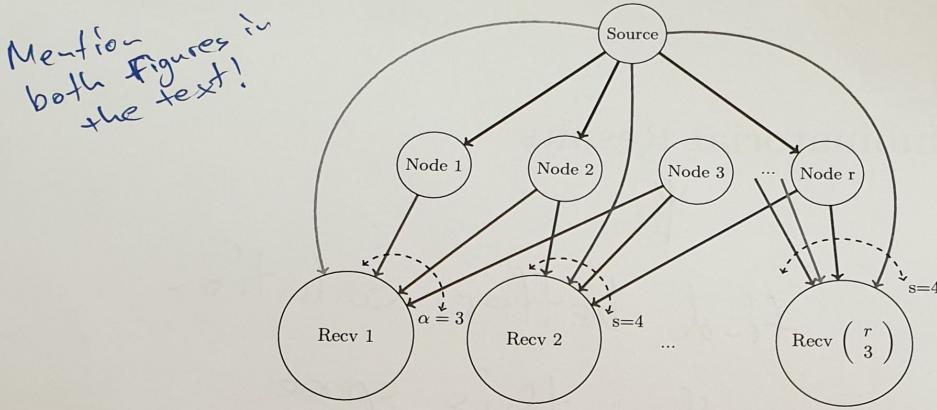
Ψ

In previous studies [EW18], no general vector solution outperforming scalar network coding was found for multicast networks with $h = 3$ messages. Hence, we start with a probabilistic argument to prove that there exists a vector solution outperforming the optimal linear solution for the $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ network. Then we generalize the proof to the $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h,r,s}$ network and the $(\epsilon = 1, \ell > 1) - \mathcal{N}_{h=2\ell,r,s=2\ell+1}$ network. As explained in Section 3.1, multiple parallel links ℓ of a data unit help us to show networks with large-capacity transmission between source and receivers. The direct links among them are not really usual in reality, i.e. a server and a client often has long-distance connection through multiple intermediate nodes, it is thus interesting to study networks with $\epsilon = 1$. We formally use r_{scalar} and r_{vector} to distinguish the r parameter of GCN for scalar solutions and vector solutions to compare their gap. Because they both have the same meaning as a number of intermediate nodes in a network, we use r when we need to state a vector solution or a scalar solution exists under some conditions of r .

5. Combinatorial Results

5.1. $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ Network

Figure 5.1.: The $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ network



In this subsection, we derive a lower bound on the maximum number of receivers for the $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ network. Due to $\alpha = 3$, the number of receivers is $N = \binom{r_{vector}}{3}$ by definition in Section 4.1. To derive the lower bound, we introduce a rank requirement on incoming packets to each receiver.

actually, on the channel matrix' rank

Figure 5.2.: The vector network coding of $(\epsilon = 1, l = 1) - \mathcal{N}_{h=3,r,s=4}$ represents as a matrix problem

Scalar Coding	Vector Coding
$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} \hline & \hline & \hline & \hline \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$	$\begin{bmatrix} \underline{y}_1 \\ \underline{y}_2 \\ \underline{y}_3 \\ \underline{y}_4 \end{bmatrix} = \begin{bmatrix} \overbrace{\hline}^{3t} & \hline & \hline & \hline \\ t & t & t & t \end{bmatrix} \begin{bmatrix} \underline{x}_1 \\ \underline{x}_2 \\ \underline{x}_3 \end{bmatrix}$
has a solution if	has a solution if
$\text{rk} \begin{bmatrix} \hline & \hline & \hline & \hline \end{bmatrix} \geq 3 \Rightarrow \text{rk} \begin{bmatrix} \hline & \hline & \hline & \hline \end{bmatrix} \geq 2$	$\text{rk} \begin{bmatrix} \overbrace{\hline}^{3t} & \hline & \hline & \hline \\ t & t & t & t \end{bmatrix} \geq 3t \Rightarrow \text{rk} \begin{bmatrix} \hline & \hline & \hline & \hline \end{bmatrix} \geq 2t$
$\Rightarrow r_{scalar} \leq 2(q_s^2 + q_s + 1)$	$\Rightarrow r_{vector} \geq ?$

5.1. ($\epsilon = 1, \ell = 1$) - $\mathcal{N}_{h=3,r,s=4}$ Network

Following to Equation 3.1, each receiver R_j must solve a linear equation system of $3t$ variables with $4t$ equations to recover $h = 3$ messages as below:

$$\begin{bmatrix} \mathbf{y}_j^{(r_1)} \\ \mathbf{y}_j^{(r_2)} \\ \mathbf{y}_j^{(r_3)} \\ \mathbf{y}_j^{(r_4)} \end{bmatrix} = \mathbf{A}_j \cdot \underline{x} = \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \\ \mathbf{A}_j^{(r_4)} \end{bmatrix} \cdot \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_3 \end{bmatrix},$$

with $\mathbf{x}_i, \mathbf{y}_j^{(v)} \in \mathbb{F}_q^t$, $\mathbf{A}_j^{(r_v)} \in \mathbb{F}_q^{t \times 3t}$ for $v = 1, \dots, 4$, and $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_3)}$ must be distinct. The network is solvable, if \mathbf{A}_j has full-rank, i.e. \mathbf{A}_{jv} must satisfy:

$$rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \\ \mathbf{A}_j^{(r_4)} \end{bmatrix} \geq 3t$$

Because coding coefficients for $\mathbf{A}_j^{(r_4)}$ can be independently chosen for any receiver R_j , there always exists $\mathbf{A}_j^{(r_4)}$ such that,

$$rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \end{bmatrix} \geq 2t \quad (5.1)$$

if and only if $rk[\mathbf{A}_j] \geq 3t$.

By this constraint, the problem is thus described as below:

$$\min_{rk[\mathbf{A}_j] \geq 3t} r_{vector}$$

$\mathbf{A}_j^{(r_1)}, \mathbf{A}_j^{(r_2)}, \mathbf{A}_j^{(r_3)}$ are matrices formed on any 3 of r links from nodes to receivers, i.e. these 3 matrices are randomly chosen from a set of r matrices. We formalize the problem by an approach with Lovász local lemma, which was initially proposed by Schwartz in [Sch].

Lemma 5.1 (Symmetric Lovász local lemma (LLL) [SCV13]). A set of events \mathcal{E}_i , with $i = 1, \dots, n$, such that each event occurs with probability at most p . If each event is independent of all others except for at most d of them and $4dp \leq 1$, then: $\Pr \left[\bigcap_{i=1}^n \overline{\mathcal{E}}_i \right] > 0$.

define event
 \mathcal{E}_i
 here

also $\mathcal{E}_{r_1, r_2, r_3}$ is a better name
 or $\mathcal{E}^{(r_1, r_2, r_3)}$

since where is i in $rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \end{bmatrix} \geq 2t$?
 The events are indexed by r_1, r_2, r_3 !

maybe use diff. letter for
 \mathcal{E}_i (LLL is very general,
 but \mathcal{E}_i is already
 the event $rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \end{bmatrix} \geq 2t$)

... between source's messages and receiver's packets. ...
 ... vector solution for GCN. Finally, we form a vector solution for GCN for our study, and we recall some instances of GCN in previous studies.
 ... approaches based on LLL. We measure the difference between a vector solution and a corresponding scalar solution for some instances of GCN in previous studies.
 ... together with Chapter 5. In Chapter 5, we also begin this chapter and the g...
 ... corollary number main the g...
 ... c...
Lemma 5.2. Let $\Pr[\mathcal{E}_i] = \Pr\left[rk\begin{bmatrix} A_j^{(r_1)} \\ A_j^{(r_2)} \\ A_j^{(r_3)} \end{bmatrix} < 2t\right]$ where $1 \leq r_1 < r_2 < r_3 \leq r$, and $A_j^{(r_1)}, \dots, A_j^{(r_3)} \in \mathbb{F}_q^{t \times 3t}$. Then, chosen independently and uniformly at random. Then,

$$\Pr[\mathcal{E}_i] \leq p \Theta(q^{-t^2-2t-1}), \forall t \geq 2. \quad (*)$$

Proof. In Lemma 5.1, each event \mathcal{E}_i is a bad event, whose occurrence is undesirable. Following to Equation 5.1, such a event occurs when $rk[A_j] < 3t$, and its probability is bounded by p (with $0 \leq p \leq 1$) as following,

$$\Pr[\mathcal{E}_i] = \Pr\left[rk\begin{bmatrix} A_j^{(r_1)} \\ A_j^{(r_2)} \\ A_j^{(r_3)} \end{bmatrix} < 2t\right] \leq p. \quad (5.2)$$

Regarding to the left-hand side,

$$\begin{aligned}
 \Pr[\mathcal{E}_i] &= \Pr\left[rk\begin{bmatrix} A_j^{(r_1)} \\ A_j^{(r_2)} \\ A_j^{(r_3)} \end{bmatrix} < 2t\right] = \sum_{i=0}^{2t-1} \Pr\left[rk\begin{bmatrix} A_j^{(r_1)} \\ A_j^{(r_2)} \\ A_j^{(r_3)} \end{bmatrix} = i\right] \\
 &\stackrel{?}{=} \sum_{i=0}^{2t-1} \frac{N_{j,m,n}}{q^{m \cdot n}} \text{ introduce notation before using if (e.g. explain it in one sentence above)}
 \end{aligned}$$

do not use m & n here! directly plug in actual matrix dimensions!

$$\begin{aligned}
 &= \sum_{i=0}^{2t-1} \prod_{j=0}^{i-1} \frac{(q^{m-j})(q^n-q^j)}{q^{i-j}} \\
 &\stackrel{?}{=} \sum_{i=0}^{2t-1} \frac{\prod_{j=0}^{i-1} (q^{3t}-q^j)^2}{q^{9t^2}} =: p. \quad (5.3)
 \end{aligned}$$

(1): The formula for the number of $[m \times n]$ matrices of rank i over \mathbb{F}_q was proved in [Ove07].

(2): $A_j^{(r_1)}, A_j^{(r_2)}, A_j^{(r_3)}$ vertically together form a $[3t \times 3t]$ matrix. In the following, we view it as a polynomial in q .

~~we consider the numerator of Equation (5.3):~~ Due to i -times product and large t :

$$\begin{cases} \deg(p_N^{(i)}(q)) = q^{i6t} \\ \deg(p_D^{(i)}(q)) = q^{i^2} \end{cases} \Rightarrow p^{(i)}(q) \approx q^{i6t-i^2}.$$

~~(*) it is wrong to say~~

26 " $a \leq p \Rightarrow p \leq \Theta(q^{-t^2-2t-1})$ "

since ~~$a \leq p$~~ also $a \leq q^{t^{9999}}$ (and obviously $q^{t^{9999}} \in \Theta(q^{-t^2-2t-1})$)
 Please note the fundamental difference of " $a \leq p \Rightarrow p \in \dots$ " and " $\text{let } \dots, \text{ then } a \leq p$ ".

5.1. ($\epsilon = 1, \ell = 1$) - $\mathcal{N}_{h=3,r,s=4}$ Network

Therefore, we have: $\sum_{i=0}^{2t-1} \prod_{j=0}^{i-1} \frac{(q^{3t} - q^j)^2}{q^i - q^j} = \sum_{i=0}^{2t-1} p^{(i)}(q) \approx \sum_{i=0}^{2t-1} q^{i6t-i^2}$.

To maximize the sum, we set derivation of it to 0 and find the corresponding root:

$$\begin{aligned} (i6t - i^2)' &= 0 \\ \Leftrightarrow 6t - 2i &= 0 \\ \Leftrightarrow i &= 3t. \end{aligned}$$

However, the upper limit of the sum is $(2t - 1)$, which is less than $3t$ for all $t \geq 2$.

$$\Rightarrow \max \left\{ q^{i6t-i^2} : i = 0, 1, \dots, 2t - 1 \right\} = q^{i6t-i^2} \Big|_{i=2t-1} = q^{8t^2-2t-1}.$$

Hence, by using the exact bound Θ , we have:

$$\max_i \left[\sum_{i=0}^{2t-1} p^{(i)}(q) \right] \in \Theta \left(\max \left\{ q^{i6t-i^2} : i = 1, 2, \dots, 2t - 1 \right\} \right) = \Theta \left(q^{8t^2-2t-1} \right) \text{ some people consider a ":" before an equation to be bad style.}$$

$$\Rightarrow \max_i \left[\frac{\sum_{i=0}^{2t-1} p^{(i)}(q)}{q^{9t^2}} \right] \in \Theta \left(q^{-t^2-2t-1} \right)$$

$$\Rightarrow p \leq \Theta \left(q^{-t^2-2t-1} \right)$$

leave out unnecessary info!

Lemma 5.3. Let $\Pr[\mathcal{E}_i] = \Pr \left[\text{rk} \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \end{bmatrix} < 2t \right] \leq p, \forall 1 \leq r_1 < r_2 < r_3 \leq r$ with $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_3)} \in \mathbb{F}_q^{t \times 3t}$, and each event \mathcal{E}_i is independent of all others except for at most d of them, then $d \leq \frac{3}{2}r^2$. why needed? just state that \mathcal{E}_i is dep. on at most $d \leq \frac{3}{2}r^2$ other events

Proof. Because d is a function of r in our problem, we denote d in Lemma 5.1 specifically by $d(r)$. The Local lemma 5.1 allows dependence among at most $d(r)$ events, i.e. each event considered under the Local lemma must no be independent to each other.

This should not be part of the proof. Furthermore, $\mathbf{A}_j^{(r_1)}, \mathbf{A}_j^{(r_2)}, \mathbf{A}_j^{(r_3)}$ are matrices formed on any 3 of r links from nodes to receivers. Assume that 1 of r links is firstly chosen to form the matrix $\mathbf{A}_j^{(r_1)}$, then there are $\binom{r-1}{2}$ possibilities left to choose $\mathbf{A}_j^{(r_2)}$ and $\mathbf{A}_j^{(r_3)}$ from $r-1$ links. However, such a link can also be chosen firstly to form $\mathbf{A}_j^{(r_2)}$ or $\mathbf{A}_j^{(r_3)}$ instead. Therefore, we have an upper (unnecessary info). Can be written as a sentence above lemma.

If $\mathcal{E}_{r_1, r_2, r_3}$ is indexed by the r_i , then²⁷ if

becomes clear that $\mathcal{E}_{r_1, r_2, r_3}$ is dependent on $\mathcal{E}_{r_1, r_2, r_3}$ if and only if $\{r_1, r_2, r_3\} \cap \{r'_1, r'_2, r'_3\} \neq \emptyset$

we introduce source's messages and receiver's responses. Why we choose GCN for our study, and ensure the difference in alphabet sizes between a vector solution and a scalar solution. Together with our new findings in Chapter 5, we begin this chapter with the gap of the network, and we also prove the number r of interest. The gap for the $(\epsilon = 1, \ell)$ computation. Chapter 6 solution with the We prove the existence of a scalar or a vector solution for GCN. Finally, gaps for some instances of GCN in previous studies. Vector approaches based on LLL. We prove of the gap, if the $\epsilon = 1 - N_{h=3,r,s=4}$ network, true?

$d(r) \leq 3 \cdot \binom{r-1}{2} = 3 \cdot \frac{(r-1)(r-2)}{2} = \frac{3}{2}(r^2 - 3r + 2)$
 $\Rightarrow d(r) \leq \frac{3}{2}r^2$

Example 5.1. $r \in \{1, 2, 3, 4\}$. Our sample space to draw any 3 matrices with orders has in total $\binom{4}{3} = 4$ samples. The context goal is not clear!

Assume the link 1 is firstly chosen to form the matrix $A_j^{(r_1)}$, we have $\binom{3}{2} = 3$ possibilities: $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix}$. The existence of 1 in $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ gives us some information. Strange word. This is only one event.

We also have 3 possibilities, when we choose the link 1 to form the matrix $A_j^{(r_2)}$: $\begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix}$

There are again 3 possibilities, when we choose the link 1 to form the matrix $A_j^{(r_3)}$: $\begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$. These are the same events.

Let's denote the event $rk \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} < 2t$ by \mathcal{E}_1 . It is clearly to see that $Pr \left[rk \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix} < 2t \mid \mathcal{E}_1 \right] =$

The bound in Lemma 5.3 is a bit loose, because we do not neglect unsuitable matrices, e.g. $\begin{bmatrix} 3 \\ 1 \\ 4 \end{bmatrix}$. In the other words, there exists at most 9 events are dependent.

>>> IS THIS EXAMPLE CONVINCING ENOUGH? WHY IS IT NOT $d(r) \leq 3 \cdot r$? That's also why.

Everywhere! be careful that $r_1 < r_2 < r_3$!

5.1. ($\epsilon = 1, \ell = 1$) - $\mathcal{N}_{h=3,r,s=4}$ Network

$$\binom{r-1}{2} = ?$$

To be exact it is:
There is an $r_0 \in \mathbb{R} (q^{t^2/2+O(t)})$ s.t. for any $r \leq r_0$ there exists ..."

Theorem 5.1. If $r \leq \Omega(q^{t^2/2+O(t)})$, then there exists a vector solution for the $(\epsilon = 1, l = 1) - \mathcal{N}_{h=3,r,s=4}$ network.

Proof. The Local lemma 5.1 shows that there is a positive probability that none of bad events occurs: $\Pr \left[\bigcap_{i=1}^n \bar{\mathcal{E}}_i \right] > 0$. ~~the?~~ ~~If $4 \cdot p \cdot d \leq 1$~~ ?

By the intersection rule, none of bad events is equivalent to an event T that its set of outcomes are all desirable, i.e. these outcomes satisfy the Equation 5.1. (S.1) :

$$T = \bigcap_{i=1}^n \bar{\mathcal{E}}_i = \left\{ \text{rk} \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \mathbf{A}_j^{(r_2)} \\ \mathbf{A}_j^{(r_3)} \end{bmatrix} \geq 2t, \forall 1 \leq r_1 < r_2 < r_3 \leq r \right\}.$$

The probability of event T indicates a measure quantifying the likelihood that we are able to construct $\text{rk}[\mathbf{A}_j] \geq 3t$ given r links. It means that a vector solution exists if and only if the Local lemma 5.1 is satisfied such that: $4 \cdot p \cdot d(r) \leq 1, \forall r \leq r_{\max, \text{vector}}$. Therefore, we must find a lower bound of $r_{\max, \text{vector}}$. Furthermore, we have $d \leq \frac{3}{2}r^2$ following to Lemma 5.3, which gives: $4 \cdot p \cdot \frac{3}{2}r^2 \leq 1 \Rightarrow r \leq \sqrt{\frac{1}{6p}} = r_{\max, \text{vector}}$. Thus, finding $\min \{r_{\max, \text{vector}}\}$ is equivalent to maximize p .

By Lemma 5.2, we have $p \leq \Theta(q^{-t^2-2t-1})$,

why min? \rightarrow use \min max
we want to \max -it! (roman letters for operators!)

sufficient condition of the

Hence, the Local lemma in 5.1 is satisfied, when $r \leq \Omega(q^{t^2/2+O(t)})$. None of bad events occurs, so there exists a vector solution for such r . \square

Corollary 5.1. The $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ network has a vector solution with a gap $q^{t^2/4+O(t)}$.

Proof.:

In [EW18, Sec. VIII-C], we have that $r_{\max, \text{scalar}} \in \mathcal{O}(q_s^2)$, where they proved that

$$r_{\text{scalar}} \leq 2 \begin{bmatrix} 3 \\ 1 \end{bmatrix}_{q_s} = 2(q_s^2 + q_s + 1). \quad (5.4)$$

5. Combinatorial Results

Following to Section 3.4.2 and Theorem 5.1, we have the gap size

$$\begin{aligned} r_{\max, \text{scalar}} &= \min \{r_{\max, \text{vector}}\} \\ \Leftrightarrow q_s^2 &= q^{t^2/2 + \mathcal{O}(t)} \\ \Leftrightarrow q_s &= q^{t^2/4 + \mathcal{O}(t)} \\ \Rightarrow g &= q_s - q_v = q^{t^2/4 + \mathcal{O}(t)} \end{aligned} \quad (5.5)$$

By varying t in Equation (5.3), we have the following table:

□

Table 5.1.: r over variations of t

t	Scalar Solution	Vector Solution
1	$r_{\text{scalar}} \leq 14$	$r_{\text{vector}} \geq 3$
2	$r_{\text{scalar}} \leq 42$	$r_{\text{vector}} \geq 7$ (67*, 89**)
3	$r_{\text{scalar}} \leq 146$	$r_{\text{vector}} \geq 62$ (166*)
4	$r_{\text{scalar}} \leq 546$	$r_{\text{vector}} \geq 1317$
5	$r_{\text{scalar}} \leq 2114$	$r_{\text{vector}} \geq 58472$
6	$r_{\text{scalar}} \leq 8322$	$r_{\text{vector}} > 10^6$

* , **: Computational results in Construction 1 and Construction 2 respectively

In the table (5.1), the vector solution outperforms the scalar solution when $t \geq 4$ for the network $(\epsilon = 1, \ell = 1) - N_{h=3, r, s=4}$. This is sufficient, we show in Section 6 computational results which vector solutions outperform scalar solutions in case of $t = 2$ and $t = 3$.

5.2. $(\epsilon = 1, \ell = 1) - N_{h,r,s}$ Network

5.2.1. Find the lower bound of $r_{\max, \text{vector}}$

As previous, $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_{h-\ell})} \in \mathbb{F}_q^{t \times ht}$ and we need to satisfy the following:

$$rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\ell})} \end{bmatrix} \geq ht - t \Leftrightarrow rk [\mathbf{A}_j] \geq (h-1)t$$

We can formulate it by the following coding problem in Grassmannian:

Find the largest set of subspaces from $\mathcal{G}_q(ht, t)$ such that any α subspaces of the set span a subspace of dimension at least $(h-1)t$.

+ this also shows that Explain everything!
+ this also shows that $r_{\max, \text{scalar}}$ or r_{scalar} ??!

Constr. 1 & 2
not introduced

Same comments as in?

with $(\epsilon = 1, \ell = 1) - N_{3,r,4}$, we consider p to proceed the Local lemma 5.1:

$$\Pr [rk [\mathbf{A}_j] < (h-1)t] \leq p \quad (5.6)$$

Lemma 5.4. Let $\Pr [\mathcal{E}_i] = \Pr \left[rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix} < (h-1)t \right] \leq p, \forall 1 \leq r_1 < \dots < r_{h-\epsilon} \leq r$ and $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_{h-\epsilon})} \in \mathbb{F}_q^{t \times ht}$, then,

$$p \leq \Theta \left(q^{(h-\alpha)t^2 + \mathcal{O}(t)} \right), \forall t \geq 2.$$

Proof. Regarding to the left-hand side in Equation 5.6:

$$\begin{aligned} \Pr [rk [\mathbf{A}_j] < (h-1)t] &= \sum_{i=0}^{(h-1)t-1} \Pr [rk [\mathbf{A}] = i] \\ &\stackrel{1}{=} \sum_{i=0}^{(h-1)t-1} \frac{N_{i,\alpha t,ht}}{q^{(\alpha t)(ht)}} \\ &= \frac{1}{q^{(\alpha h)t^2}} \cdot \sum_{i=0}^{(h-1)t-1} \prod_{j=0}^{i-1} \frac{(q^{\alpha t} - q^j)(q^{ht} - q^j)}{q^i - q^j}. \end{aligned} \quad (5.7)$$

(1): The formula for the number of $[m \times n]$ matrices of rank i over \mathbb{F}_q was proved in [Ove07].

We consider firstly the product $\prod_{j=0}^{i-1} \frac{(q^{\alpha t} - q^j)(q^{ht} - q^j)}{q^i - q^j} = \frac{p_N^{(i)}(q)}{p_D^{(i)}(q)} = p^{(i)}(q)$.

$$\text{For } t \rightarrow \infty: \left. \begin{array}{l} \deg(p_N^{(i)}(q)) = q^{i(\alpha t + ht)} \\ \deg(p_D^{(i)}(q)) = q^{i^2} \end{array} \right\} \Rightarrow p^{(i)}(q) \approx q^{i(\alpha t + ht) - i^2}.$$

Now, we evaluate $f(i) = i(\alpha t + ht) - i^2$ to find its maximum point by its derivation :

$$f'(i^*) = 0 \Leftrightarrow (\alpha t + ht) - 2i^* = 0 \Leftrightarrow i^* = \frac{\alpha t + ht}{2}$$

We then check whether this point within the range $i = 0, \dots, (h-1)t-1$ as following:
 $0 \leq \frac{\alpha t + ht}{2} \leq (h-1)t-1$.

With regards to the lower bound: $0 \leq \frac{\alpha t + ht}{2} \Leftrightarrow t \geq \frac{2}{\alpha+h}$, which is always true due to the given $t \geq 2$ and $\alpha, h \geq 3$.

Regarding to the upper bound: $\frac{\alpha t + ht}{2} \leq (h-1)t-1 \Leftrightarrow t \leq \frac{-2}{\alpha+2-h}$ with $\alpha+2 > h$ due to the given $\alpha\ell + \epsilon = \alpha + 1 \geq h$. This cannot happen because of $t \geq 2$, i.e. this maximum

5. Combinatorial Results

point is over then upper-range limit.

$$\Rightarrow \max \left\{ q^{i(\alpha t + ht) - i^2} : i = 1, \dots, (h-1)t-1 \right\} = q^{i(\alpha t + ht) - i^2} \Big|_{i=(h-1)t-1} = q^{[(h-1)(\alpha+1)]t^2 - (\alpha-h+2)t - 1}.$$

Secondly, we apply the maximum value with the sum, we have:

$$\begin{aligned} \max \left\{ \sum_{i=0}^{(h-1)t-1} p^{(i)}(q) \right\} &\in \Theta \left(q^{[(h-1)(\alpha+1)]t^2 + O(t)} \right) \\ \Rightarrow \max \left\{ \frac{\sum_{i=0}^{(h-1)t-1} p^{(i)}(q)}{q^{(\alpha h)t^2}} \right\} &\in \left(\frac{q^{[(h-1)(\alpha+1)]t^2 + O(t)}}{q^{(\alpha h)t^2}} \right) \\ \Rightarrow p &\leq \Theta \left(q^{(h-\alpha)t^2 + O(t)} \right) \end{aligned}$$

□

Lemma 5.5. Let $\Pr[\mathcal{E}_i] = \Pr \left[rk \begin{bmatrix} A_j^{(r_1)} \\ \vdots \\ A_j^{(r_{h-\epsilon})} \end{bmatrix} < (h-1)t \right] \leq p, \forall 1 \leq r_1 < \dots < r_{h-\epsilon} \leq r$ and $A_j^{(r_1)}, \dots, A_j^{(r_{h-\epsilon})} \in \mathbb{F}_q^{txht}$, and each event \mathcal{E}_i is independent of all others except for at most d of them, then $d \leq \frac{\alpha}{(\alpha-1)!} r^{\alpha-1}$.

Proof. Similar with $(\epsilon = 1, \ell = 1) - N_{3,r,4}$, we have:

$$d \leq \alpha \binom{r-1}{\alpha-1} = \alpha \frac{(r-1)\dots(r-\alpha+1)}{(\alpha-1)!} \leq \frac{\alpha}{(\alpha-1)!} r^{\alpha-1}$$

□

Theorem 5.2. If $r \leq \Omega \left(q^{\frac{h-\alpha-1}{1-\alpha} t^2 + O(t)} \right)$, then there exists a vector solution for the $(\epsilon = 1, \ell = 1) - N_{h,r,s}$ network.

Proof. As previous, satisfying LLL 5.1 is equivalent to an existence of a vector solution. We need $4dp \leq 1$, and following to Lemma 5.5: $d \leq \frac{\alpha}{(\alpha-1)!} r^{\alpha-1} \Rightarrow r \leq \left(\frac{(\alpha-1)!}{4\alpha} \cdot \frac{1}{p} \right)^{\frac{1}{\alpha-1}} = r_{max, vector}$. We thus again find a lower bound of $r_{max, vector}$.

By Lemma 5.5, we have $p \leq \Theta \left(q^{(h-\alpha)t^2 + O(t)} \right), \forall t \geq 2$,

$$\Rightarrow \min \{r_{max, vector}\} \in \Omega \left(q^{\frac{h-\alpha-1}{1-\alpha} t^2 + O(t)} \right).$$

Hence, the Local lemma 5.1 is satisfied, when $r \leq \Omega \left(q^{\frac{h-\alpha-1}{1-\alpha} t^2 + O(t)} \right)$, and a vector solution exists for such r .

5.3. $(\epsilon = 1, \ell \geq 2) - \mathcal{N}_{h=2\ell, r,s=2\ell+1}$

5.2.2. Find the Upper Bound of $r_{max,scalar}$
 There is no unique upper bound

Find $(\alpha + 1)$ received vectors that span a subspace of dimension h . This implies that the α links from the middle layer carry α vectors which span a subspace of $\mathbb{F}_{q_s}^h$ whose dimension is at least $(h - 1)$, with $q_s = q^t$.

Following to Theorem 4.1, we are interested in the following range: $\ell + \epsilon + 1 \leq h \leq \alpha\ell + \epsilon$.
 (everywhere!)

For $3 \leq \alpha < h$, all α links must be distinct $\Rightarrow r \leq \begin{bmatrix} \alpha \\ 1 \end{bmatrix}_{q_s} \Rightarrow r \leq \mathcal{O}(q_s^{\alpha-1})$

For $\alpha \geq h \geq 3$: to achieve $(h - 1)$ -subspaces of $\mathbb{F}_{q_s}^h$, no α links will contain a vector which is contained in the same $(h - 2)$ -subspace.

Hence,

$$r_{max,scalar} \leq (\alpha - 1) \begin{bmatrix} \alpha \\ h - 2 \end{bmatrix}_{q_s} \Rightarrow r_{max,scalar} \in \mathcal{O}(q_s^{(\alpha-h+2)(h-2)t^2})$$

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5.2.3. Calculate Gap

+ a sentence

$$\begin{aligned} r_{max,scalar} &= \min \{r_{max,vector}\} \\ \Leftrightarrow q_s^{(\alpha-h+2)(h-2)t^2} &= q^{\frac{h-\alpha-1}{1-\alpha}t^2 + \mathcal{O}(t)} \\ \Leftrightarrow q_s &= q^{\frac{\alpha-h+1}{(\alpha-1)(\alpha-h+2)(h-2)}t^2 + \mathcal{O}(t)} \\ \Rightarrow g &= q_s - q_v = q^{\frac{\alpha-h+1}{(\alpha-1)(\alpha-h+2)(h-2)}t^2 + \mathcal{O}(t)} \end{aligned}$$

state as
Thm.

5.3. $(\epsilon = 1, \ell \geq 2) - \mathcal{N}_{h=2\ell, r,s=2\ell+1}$

+ Text

Lemma 5.6. Let $Pr[\mathcal{E}_i] = Pr \left[rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix} < (2\ell - 1)t \right] \leq p, \forall 1 \leq r_1 < \dots <$

$r_{h-\epsilon} \leq r$ and $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_{h-\epsilon})} \in \mathbb{F}_q^{t \times 2\ell t}$, then,

$$p \leq \Theta(q^{-t^2 - 2t - 1}), \forall t \geq 2.$$

Same comments as Sec. 5.1

carried out
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(-12)

Local Lemma (LLL), and Chapter 6 and 7 contains new computational results of vector solutions outperforming scalar solutions for the $(\epsilon = 1, \ell = 1) - \mathcal{N}_{h=3,r,s=4}$ network. The details of each chapter are mentioned below.

In **Chapter 3** and **Chapter 4**, we represent network as a matrix channel and introduce an advantage of vector solutions in alphabet sizes in comparison with scalar solutions for network coding problems. We firstly recall the motivation of network coding, and secondly we introduce how we approach such problems by formulating the relationship between source's messages and receiver's packets by linear equation systems. Thirdly, we explain

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Hence, by using the exact bound Θ , we have:
$$\max_{i=0}^{(2\ell-1)t-1} q^{i4\ell t - i^2}$$

5. Combinatorial Results

Proof. A bad event in this vector network coding problem has the following probability:

$$\begin{aligned} \Pr \left[rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix} < (2\ell-1)t \right] &= \sum_{i=0}^{(2\ell-1)t-1} \Pr \left[rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix} = i \right] \\ &\stackrel{1}{=} \sum_{i=0}^{(2\ell-1)t-1} \frac{N_{i,m,n}}{q^{m \cdot n}} \\ &\stackrel{2}{=} \frac{1}{q^{4\ell^2 t^2}} \sum_{i=0}^{(2\ell-1)t-1} \prod_{j=0}^{i-1} \frac{(q^{2\ell t} - q^j)^2}{q^i - q^j} \end{aligned} \quad (5.8)$$

(1): The formula for the number of $[m \times n]$ matrices of rank i over \mathbb{F}_q was proved in [Ove07].

(2): $s = \alpha\ell + \epsilon$ by definition in Section 4.1 $\Rightarrow \alpha = 2$, so $\mathbf{A}_j \in \mathbb{F}_q^{2\ell t \times 2\ell t}$ with $\mathbf{A}_j = \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix}$, and \mathbf{A}_j contains (2ℓ) t -dimensional subspaces of $\mathbb{F}_q^{2\ell t}$.

We consider the product in Equation 5.8: $\prod_{j=0}^{i-1} \frac{(q^{2\ell t} - q^j)^2}{q^i - q^j} = \frac{p_N^{(i)}(q)}{p_D^{(i)}(q)} = p^{(i)}(q)$.

Due to i -times product and large t :
$$\begin{cases} \deg(p_N^{(i)}(q)) = q^{i4\ell t} \\ \deg(p_D^{(i)}(q)) = q^{i^2} \end{cases} \Rightarrow p^{(i)}(q) \approx q^{i4\ell t - i^2}.$$

Therefore, we have: $\sum_{i=0}^{(2\ell-1)t-1} \prod_{j=0}^{i-1} \frac{(q^{2\ell t} - q^j)^2}{q^i - q^j} = \sum_{i=0}^{(2\ell-1)t-1} p^{(i)}(q) \approx \sum_{i=0}^{(2\ell-1)t-1} q^{i4\ell t - i^2}$.

To maximize the sum, we set derivation of it to 0 and find the corresponding root:

$$\begin{aligned} (i4\ell t - i^2)' &= 0 \\ \Leftrightarrow 4\ell t - 2i &= 0 \\ \Leftrightarrow i &= 2\ell t \end{aligned}$$

However, the upper limit of the sum is $(2\ell - 1)t - 1$, which is less than $2\ell t$ for all $t \geq 2$.

$$\Rightarrow \max \left\{ q^{i4\ell t - i^2} : i = 0, 1, \dots, (2\ell - 1)t - 1 \right\} = q^{i4\ell t - i^2} \Big|_{i=(2\ell-1)t-1} = q^{4\ell^2 t^2 - t^2 - 2t - 1}$$



 5.3. $(\epsilon = 1, \ell \geq 2) - \mathcal{N}_{h=2\ell, r,s=2\ell+1}$

Hence, by using the exact bound Θ , we have:

$$\begin{aligned} \max_i \left\{ \sum_{i=0}^{(2\ell-1)t-1} p^{(i)}(q) \right\} &\in \Theta(q^{4\ell^2t^2-t^2-2t-1}) \\ \Rightarrow \max_i \left\{ \frac{1}{q^{4\ell^2t^2}} \sum_{i=0}^{(2\ell-1)t-1} p^{(i)}(q) \right\} &\in \Theta(q^{-t^2-2t-1}) \\ \Rightarrow p &\leq \Theta(q^{-t^2-2t-1}) \end{aligned}$$

□

Lemma 5.7. Let $\Pr[\mathcal{E}_i] = \Pr[rk \begin{bmatrix} \mathbf{A}_j^{(r_1)} \\ \vdots \\ \mathbf{A}_j^{(r_{h-\epsilon})} \end{bmatrix} < (2\ell-1)t] \leq p, \forall 1 \leq r_1 < \dots < r_{h-\epsilon} \leq r$ and $\mathbf{A}_j^{(r_1)}, \dots, \mathbf{A}_j^{(r_{h-\epsilon})} \in \mathbb{F}_q^{t \times 2\ell t}$, and each event \mathcal{E}_i is independent of all others except for at most d of them, then $d \leq 2r$.

Proof. Being similar with the previous subsections, we have:

$$d \leq \alpha \binom{r-1}{\alpha-1} = 2 \frac{(r-1)\dots(r-1)}{1!} \leq 2r$$

□

Theorem 5.3. If $r \leq \Omega(q^{t^2+\mathcal{O}(t)})$, then there exists a vector solution for the $(\epsilon = 1, \ell \geq 2) - \mathcal{N}_{h=2\ell, r,s=2\ell+1}$ network.

Proof. As previous, we need $4 \cdot p \cdot d(r) \leq 1, \forall r \leq r_{max, vector}$ so that a vector solution exists. Following to Lemma 5.7, we have $d \leq 2r \Rightarrow 4 \cdot p \cdot 2r \leq 1 \Rightarrow r \leq \frac{1}{8p}$. We still need to maximize p to get lower on $r_{max, vector}$, and we have $p \leq \Theta(q^{-t^2-2t-1}), \forall t \geq 2$ in Lemma 5.6. Thus, $\min \{r_{max, vector}\} \in \Omega\left(\frac{1}{8p}\right) = \Omega(q^{t^2+2t+1})$.

Hence, the Local lemma in 5.1 is satisfied, when $r \leq \Omega(q^{t^2/2+\mathcal{O}(t)})$. None of bad events occurs, so there exists a vector solution for such r . □

Lemma 5.8. A scalar solution for the $(\epsilon = 1, \ell \geq 2) - \mathcal{N}_{h=2\ell, r,s=2\ell+1}$ network exists, if and only if there exists a Grassmannian code $\mathcal{G}_q(h = 2\ell, \ell)$ such that any $\alpha = 2$ subspaces of the set span a subspace of dimension at least $2\ell - 1$.

Proof. Let's denote any 2 subspaces of $\mathcal{G}_q(h = 2\ell, \ell)$ as \mathcal{U} and \mathcal{V} . If \mathcal{U} and \mathcal{V} span a subspace of dimension at least $2\ell - 1$, then we have $\dim(\mathcal{U} + \mathcal{V}) = 2\ell - 1$, with $\mathcal{U} + \mathcal{V}$.

Therefore, $\dim(\mathcal{U} \cap \mathcal{V}) = \dim(\mathcal{U}) + \dim(\mathcal{V}) - \dim(\mathcal{U} + \mathcal{V}) = 1$, which leads to the subspace distance $d_s(\mathcal{U}, \mathcal{V}) = 2\dim(\mathcal{U} + \mathcal{V}) - \dim(\mathcal{U}) - \dim(\mathcal{V}) = 2\ell - 2$. No 2ℓ -dimensional subspaces of $\mathbb{F}_{q_s}^{2\ell}$ will contain a vector which is contained in the same $(2\ell - 2)$ -subspace, but $(2\ell - 1)$ of such subspaces can have such vectors,

$$\Rightarrow r_{scalar} \leq (2\ell - 1) \begin{bmatrix} 2\ell \\ 2\ell - 2 \end{bmatrix}_{q_s} \Rightarrow r_{scalar} \leq \mathcal{O}(q_s^{2\ell-1})$$

□

Corollary 5.2. *The ($\epsilon = 1, \ell \geq 2$) - $\mathcal{N}_{h=2\ell, r, s=2\ell+1}$ network has a vector solution with a gap $q^{t^2/2\ell + \mathcal{O}(t)}$.*

Following to Section 3.4.2 and Theorem 5.3, we have the gap size

$$\begin{aligned} r_{max, scalar} &= \min \{r_{max, vector}\} \\ \Leftrightarrow q_s^\ell &= q^{t^2/2 + \mathcal{O}(t)} \\ \Leftrightarrow q_s &= q^{t^2/2\ell + \mathcal{O}(t)} \\ \Rightarrow g &= q_s - q_v = q^{t^2/2\ell + \mathcal{O}(t)} \end{aligned} \tag{5.9}$$

This shows us that there exists a better vector solution by comparison with the gap in [EW18, Fig. 4].