

# Constant Approximating $k$ -Clique is $W[1]$ -hard

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## Abstract

For every graph  $G$ , let  $\omega(G)$  be the largest size of complete subgraph in  $G$ . This paper presents a simple algorithm which, on input a graph  $G$ , a positive integer  $k$  and a small constant  $\epsilon > 0$ , outputs a graph  $G'$  and an integer  $k'$  in  $2^{\Theta(k^5)} \cdot |G|^{O(1)}$ -time such that (1)  $k' \leq 2^{\Theta(k^5)}$ , (2) if  $\omega(G) \geq k$ , then  $\omega(G') \geq k'$ , (3) if  $\omega(G) < k$ , then  $\omega(G') < (1 - \epsilon)k'$ . This implies that no  $f(k) \cdot |G|^{O(1)}$ -time algorithm can distinguish between the cases  $\omega(G) \geq k$  and  $\omega(G) < k/c$  for any constant  $c \geq 1$  and computable function  $f$ , unless  $FPT = W[1]$ .

## 1 Introduction

Given a simple graph  $G$  and a positive integer  $k$ , the task of  $k$ -Clique problem is to decide whether  $\omega(G) \geq k$ . In parameterized complexity [DF99, FG06], the  $k$ -Clique problem with  $k$  as its parameter is a canonical  $W[1]$ -complete problem [DF95]. Unless  $W[1] = FPT$ , it has no  $f(k) \cdot |G|^{O(1)}$ -time algorithm (FPT-algorithm) for any computable function  $f : \mathbb{N} \rightarrow \mathbb{N}$ . This problem has been used as a starting point in many reductions and thus plays a fundamental role in the area of parameterized complexity. Yet, it is still not known whether constant approximating  $k$ -Clique is also  $W[1]$ -hard. More precisely, we consider the following question:

**Question 1.1.** *Is there any algorithm which, on input a graph  $G$  and a positive integer  $k$ , outputs a new graph  $G'$  and a positive integer  $k'$  in  $f(k) \cdot |G|^{O(1)}$ -time for some computable function  $f : \mathbb{N} \rightarrow \mathbb{N}$  such that,*

- $k' = g(k)$  for some computable function  $g : \mathbb{N} \rightarrow \mathbb{N}$ ,
- if  $\omega(G) \geq k$ , then  $\omega(G') \geq k'$ ,
- if  $\omega(G) < k$ , then  $\omega(G') < k'/2$ ?

The question above is motivated by the study of FPT-approximation algorithms for  $k$ -Clique. For any  $c \geq 1$ , we say an algorithm is a  $c$ -FPT-approximation algorithm for  $k$ -Clique if on input a graph  $G$  it outputs a clique of size  $\omega(G)/c$  in  $G$  in  $f(\omega(G)) \cdot |G|^{O(1)}$ -time for some computable function  $f : \mathbb{N} \rightarrow \mathbb{N}$ . Whether there exists such an algorithm has been repeatedly raised in the literature [Mar08, FGMS, CGG06, DF13]. Previous results for FPT-inapproximability of  $k$ -Clique are under strong assumptions which already have a gap [CCK<sup>+</sup>17, BEKP15]. Proving such results based on standard assumptions is an interesting and important open question. It is wildly believed that the technique needed to resolve this question is closely related to a PCP-theorem for parameterized complexity [CGG06]:

Non-approximability results in the classical framework were proved for the CLIQUE problem using the PCP-theorem, so it might be necessary to obtain a parameterized version of the PCP-theorem to solve these questions.

This paper gives the first positive answer to Question 1.1. An immediate corollary of our result is the non-existence of FPT-approximation algorithm for the  $k$ -Clique problem under the standard parameterized complexity hypothesis  $W[1] \neq FPT$ .

**Theorem 1.2.** *Assuming that  $k$ -Clique has no FPT-algorithm, there is no FPT-algorithm that can approximate  $k$ -Clique to any constant.*

The main contribution of this paper is to show how to create a constant gap for  $\omega(G)$  from a  $W[1]$ -hard problem with no gap.

## 1.1 Overview of the reduction

Let us illustrate the idea using a toy  $k$ -Vector-Sum problem and the Walsh-Hadamard code. Throughout this paper, we will work on finite field  $\mathbb{F}$  with characteristic 2. For any  $\vec{v}_1, \dots, \vec{v}_k \in \mathbb{F}^d$  and  $\vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^d$ , let  $H(\vec{v}_1, \dots, \vec{v}_k)_{\vec{a}_1, \dots, \vec{a}_k} = \sum_{i \in [k]} \vec{a}_i \cdot \vec{v}_i$  be the Walsh-Hadamard code of  $\vec{v}_1, \dots, \vec{v}_k$ . Here  $\vec{a}_i \cdot \vec{v}_i$  denotes the dot product of vectors  $\vec{a}_i$  and  $\vec{v}_i$ . Given  $k$  vector  $\vec{v}_1, \dots, \vec{v}_k \in \mathbb{F}^d$  and a target vector  $\vec{t} \in \mathbb{F}^d$ , we want to test whether  $\sum_{i \in [k]} \vec{v}_i = \vec{t}$ . For that sake, we construct a constraint satisfaction problem (CSP) on variables  $\{x_{\vec{a}_1, \dots, \vec{a}_k} : \vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^d\}$ . Any assignment to these variables can be seen as a long vector  $\vec{x} \in \mathbb{F}^{kd}$ , which is supposed to be the Walsh-Hadamard code of  $\vec{v}_1, \dots, \vec{v}_k$ . Then we do the following tests.

- (T1) To ensure that the vector  $\vec{x} \in \mathbb{F}^{kd}$  is a Walsh-Hadamard code of some vector  $\vec{v}_1, \dots, \vec{v}_k \in \mathbb{F}^d$ , we check if  $\vec{x}_{\vec{a}_1 + \vec{b}_1, \dots, \vec{a}_k + \vec{b}_k} = \vec{x}_{\vec{a}_1, \dots, \vec{a}_k} + \vec{x}_{\vec{b}_1, \dots, \vec{b}_k}$  for  $\vec{a}_1, \dots, \vec{a}_k, \vec{b}_1, \dots, \vec{b}_k \in \mathbb{F}^d$ .
- (T2) Check if  $\vec{x}_{\vec{a}_1, \dots, \vec{a}_i + \vec{a}, \dots, \vec{a}_k} - \vec{x}_{\vec{a}_1, \dots, \vec{a}_k} = \vec{v}_i \cdot \vec{a}$  for  $i \in [k]$ ,  $\vec{a} \in \mathbb{F}^d$  and  $\vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^d$ .
- (T3) Check if  $\vec{x}_{\vec{a}_1 + \vec{a}, \dots, \vec{a}_k + \vec{a}} - \vec{x}_{\vec{a}_1, \dots, \vec{a}_k} = \vec{a} \cdot \vec{t}$  for  $\vec{a}, \vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^d$ .

By the linearity testing [BLR93], for any assignment  $\vec{x} \in \mathbb{F}^{kd}$ , if  $\vec{x}$  satisfies  $\delta$ -fractions of constraints in (T1), then  $\vec{x}$  is at most  $(1 - \delta)$ -far from the Walsh-Hadamard code of some vectors  $\vec{u}_1, \dots, \vec{u}_k \in \mathbb{F}^d$ . That is, for a  $\delta$ -fraction of  $(\vec{a}_1, \dots, \vec{a}_k) \in \mathbb{F}^{dk}$ ,  $\vec{x}_{\vec{a}_1, \dots, \vec{a}_k} = H(\vec{u}_1, \dots, \vec{u}_k)_{\vec{a}_1, \dots, \vec{a}_k}$ . Next, if  $\vec{u}_1 + \dots + \vec{u}_k \neq \vec{t}$ , then for a  $(1 - 1/|\mathbb{F}|)$ -fraction of  $\vec{a} \in \mathbb{F}^d$ ,  $\vec{a} \cdot (\vec{u}_1 + \dots + \vec{u}_k) \neq \vec{a} \cdot \vec{t}$ . If for some  $i \in [k]$ ,  $\vec{u}_i \neq \vec{v}_i$ , then for a  $(1 - 1/|\mathbb{F}|)$ -fraction of  $\vec{a} \in \mathbb{F}^d$ ,  $\vec{a} \cdot \vec{u}_i \neq \vec{a} \cdot \vec{v}_i$ . We conclude that, if  $\vec{v}_1 + \dots + \vec{v}_k = \vec{t}$ , then there exists a vector  $\vec{x}$  satisfying all the constraints. If  $\vec{v}_1 + \dots + \vec{v}_k \neq \vec{t}$ , then for any assignment  $\vec{x}$ , one of the following must hold:

- a constant fraction of constraints in (T1) are not satisfied.
- for some  $i \in [k]$ , a constant fraction of constraints in (T2) with respect to  $i$  are not satisfied.
- a constant fraction of constraints in (T3) are not satisfied.

Then we construct a graph  $G$  from the CSP instance using a modified FGLSS-reduction [FGL<sup>+</sup>96]. The vertex set of  $G$  consists of two parts  $A$  and  $B$ .

The vertices of  $A$  are corresponding to assignments to variables  $\{x_{\vec{a}_1, \dots, \vec{a}_k} : \vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^d\}$  of the CSP instance. Two vertices in  $A$  are adjacent unless they are corresponding to assignments that are not consistent, i.e., assigning different values to the same variable, or there exists a (T2) or (T3) constraint between them and they do not satisfy that constraint. Note that there are  $|\mathbb{F}|^{1+dk}$  vertices in  $A$ . They can be partitioned into  $|\mathbb{F}|^{dk}$  independent sets of size  $|\mathbb{F}|$ .

The vertices in  $B$  are corresponding to assignments to three variables satisfying constraints in (T1). Two vertices in  $B$  are adjacent unless they are corresponding to inconsistent assignments. Note that the vertex set of  $B$  can be partitioned into  $|\mathbb{F}|^{2dk}$  disjoint subsets, each forms an independent set of size  $|\mathbb{F}|^2$ .

We add an edge between any two vertices in  $A$  and  $B$  if the assignments corresponding to these vertex are consistent. Since the sizes of  $A$  and  $B$  are not balanced, we assign to each vertex in  $A$  weight  $|\mathbb{F}|^{kd}$  and each vertex in  $B$  weight 1.

If  $\vec{v}_1 + \dots + \vec{v}_k = \vec{t}$ , then  $G$  contains a clique of weight  $2|\mathbb{F}|^{2kd}$ . This clique consists of  $|\mathbb{F}|^{kd}$  vertices from  $A$  and  $|\mathbb{F}|^{2kd}$  vertices from  $B$ . If  $\vec{v}_1 + \dots + \vec{v}_k \neq \vec{t}$ , then for any clique  $X$  in  $G$ , one of the following must hold:

- (i)  $|X \cap B| \leq (1 - \epsilon)|\mathbb{F}|^{2kd}$ ,
- (ii) there exists  $\vec{a} \in \mathbb{F}^d$  and  $i \in [k]$  such that the variable set  $V$  can be partitioned into two disjoint sets  $V = V_0 \cup V_1$  with  $V_1 = \{x_{\vec{a}_1, \dots, \vec{a}_i + \vec{a}, \dots, \vec{a}_k} : x_{\vec{a}_1, \dots, \vec{a}_k} \in V_0\}$ . The assignment corresponding to  $X$  satisfies only a  $(1 - \epsilon)$ -fraction of (T2) constraints between  $V_0$  and  $V_1$ ,
- (iii) there exists  $\vec{a} \in \mathbb{F}^d$  such that the variable set  $V$  can be partitioned into two disjoint sets  $V = V_0 \cup V_1$  with  $V_1 = \{x_{\vec{a}_1 + \vec{a}, \dots, \vec{a}_k + \vec{a}} : x_{\vec{a}_1, \dots, \vec{a}_k} \in V_0\}$ . The assignment corresponding to  $X$  satisfies only a  $(1 - \epsilon)$ -fraction of (T3) constraints between  $V_0$  and  $V_1$ .

Either (ii) or (iii) implies that  $|X \cap A| \leq [1/2 + (1 - \epsilon)1/2]|\mathbb{F}|^{kd}$ . So, in summary, when  $\vec{v}_1 + \dots + \vec{v}_k \neq \vec{t}$ , we have for every clique of  $G$ , either  $|X \cap B| \leq (1 - \epsilon)|\mathbb{F}|^{2kd}$  or  $|X \cap A| \leq (1 - \epsilon/2)|\mathbb{F}|^{kd}$ . In both cases, every clique in  $G$  has at most  $(2 - \epsilon/2)|\mathbb{F}|^{2kd}$  weight.

There are two problems needed to be solved:

- (P1) In the real  $k$ -Vector-Sum problem, we are given  $k$  sets  $V_1, \dots, V_k$  of vectors instead of  $k$  vectors. How to test whether  $\vec{x}_{\vec{a}_1, \dots, \vec{a}_i + \vec{a}, \dots, \vec{a}_k} - \vec{x}_{\vec{a}_1, \dots, \vec{a}_k} = \vec{v}_i \cdot \vec{a}$  for the same vector  $\vec{v}_i \in V_i$ ? In some bad scenario, it is possible that for each  $\vec{a}$ , there exists a vector  $\vec{v}_{\vec{a}} \in V_i$  such that  $\vec{x}_{\vec{a}_1, \dots, \vec{a}_i + \vec{a}, \dots, \vec{a}_k} - \vec{x}_{\vec{a}_1, \dots, \vec{a}_k} = \vec{v}_{\vec{a}} \cdot \vec{a}$ .
- (P2) The  $k$ -Vector-Sum problem is  $W[1]$ -hard when  $d = k^{\Omega(1)} \log n$ . Applying our reduction directly would take at least  $|\mathbb{F}|^{kd} \geq n^{k^{\Omega(1)}}$  time, which we cannot afford.

We handle these problems by sampling  $\ell$  matrices  $A_1, \dots, A_\ell \in \mathbb{F}^{h \times d}$  with  $h = k^2$ , and replacing each vector  $\vec{v} \in V_i$  by an  $\ell$ -tuple  $(A_1 \vec{v}, \dots, A_\ell \vec{v})$  of  $h$ -dimension vectors. For every  $\vec{a}_1, \dots, \vec{a}_k \in \mathbb{F}^h$ , the value of the variable  $x_{\vec{a}_1, \dots, \vec{a}_k}$  becomes a vector in  $\mathbb{F}^\ell$  instead of just an element of  $\mathbb{F}$ . The number of variables become  $|\mathbb{F}|^{k^2 h}$  instead of  $|\mathbb{F}|^{kd}$ . Since  $|\mathbb{F}| = O(1)$ , the reduction can be done in FPT-time if  $\ell \leq O(\log n + h)$ . To ensure that the constraints in (T2) still work, we show that when  $\ell \geq \Omega(\log n + h)$ , with high probability, for all  $\vec{a} \in \mathbb{F}^h$  and distinct  $\vec{v}, \vec{u} \in V_i$ ,  $(\vec{a}^T A_1 \vec{v}, \dots, \vec{a}^T A_\ell \vec{v}) \neq (\vec{a}^T A_1 \vec{u}, \dots, \vec{a}^T A_\ell \vec{u})$  and the bad scenario in (P1) will not occur. For constraints in (T3), we show that with high probability, for all distinct  $\vec{t}, \vec{t}' \in \mathbb{F}^d$ ,  $(A_1 \vec{t}, \dots, A_\ell \vec{t}) \neq (A_1 \vec{t}', \dots, A_\ell \vec{t}')$ .

## 1.2 Related work

The  $k$ -Clique problem is one of the first known NP-hard problems in [Kar72]. It was showned that approximating  $k$ -Clique to a factor of  $n^{1-\epsilon}$  is also NP-hard after a long line of research [FGL<sup>+</sup>96, BGLR93, BS94, Gol98, FK00, Has96, Zuc06]. As pointed out in [CL19], the classical inapproximability results of  $k$ -Clique inevitably produce instances with large  $\omega(G)$ . However, in parameterized complexity, we consider instances with small  $\omega(G)$  which does not depend on the size of  $G$ . In order to show that  $k$ -Clique is still hard to approximate when  $k$  is small, a natural idea is to use the PCP-theorem [ALM<sup>+</sup>98, AS98] to obtain a gap for the SAT problem and then use the method of compressing to reduce the optimum solution size [HKK13]. Unfortunately, since the PCP-theorem causes a polylogarithmic blow-up in the size of SAT instance, this approach cannot rule out FPT-approximation for  $k$ -Clique.

To circumvent this, researches used stronger hypothesis to obtain a gap for the SAT problem. Bonnet et al. [BEKP15] used ETH [IPZ01] and the linear PCP conjecture to show constant FPT-inapproximability of  $k$ -Clique. Assuming Gap-ETH [Din16, MR16], it was shown that there is no  $o(k)$ -FPT-approximation for  $k$ -Clique [CCK<sup>+</sup>17].

In [LRSZ20], a weaker conjecture called Parameterized Inapproximability Hypothesis (PIH) was postulated. PIH states that binary CSP parameterized by the number of variables has no constant FPT-approximation. It is easy to see that PIH implies  $k$ -Clique has no constant FPT-approximation. Interestingly, it is not known if the hardness of approximation of  $k$ -Clique implies PIH [FKLM20].

Assuming a conjecture called DEG-2-SAT, Khot and Shinkar [KS16] used a different approach to rule out FPT-approximation for  $k$ -Clique. Although the conjecture turned out to be

false [Kay14], their work is still inspiring. The idea of multiplying the input instance with matrices in our reduction is from their paper.

In recent years, several gap-creating techniques have been successfully used to show FPT inapproximabilities [Lin18, LKM19, Lin19, Wlo20, KLN21]. We refer the reader to [FKLM20] for a survey of these results.

## 2 Preliminaries

For every vector  $\vec{v} = (v_1, \dots, v_d) \in \mathbb{F}^d$  and  $i \in [d]$ , let  $\vec{v}[i] = v_i$ . For every  $\vec{x}, \vec{y} \in \mathbb{F}^n$ , we use  $\vec{x} \cdot \vec{y} = \sum_{i \in [n]} \vec{x}[i] \vec{y}[i]$  to denote the dot product of  $\vec{x}$  and  $\vec{y}$ . For  $X \subseteq \mathbb{F}^d$  and  $\vec{v} \in \mathbb{F}^d$ , let  $X + \vec{v} = \{\vec{x} + \vec{v} : \vec{x} \in X\}$ . For  $\vec{a} \in \mathbb{F}^n$  and  $\vec{b} \in \mathbb{F}^m$ , let  $\vec{a} \circ \vec{b}$  be the result of concatenating  $\vec{a}$  and  $\vec{b}$ . For any  $dn$  vectors  $\vec{v} = (v_1, \dots, v_{dn}) \in \mathbb{F}^{dn}$  and  $\vec{a} = (a_1, \dots, a_d) \in \mathbb{F}^d$ , let

$$F(\vec{a}, \vec{v}) = \left( \sum_{i \in [d]} a_i v_i, \dots, \sum_{i \in [d]} a_i v_{i+jd}, \dots, \sum_{i \in [d]} a_i v_{i+(n-1)d} \right) \in \mathbb{F}^n.$$

For all  $n \in \mathbb{N}$ , let  $\vec{0}_n$  and  $\vec{1}_n$  be the  $n$ -dimension all-zero vector and all-one vector respectively.

**Definition 2.1** (Distance). *For every  $d \in \mathbb{N}$  and  $\vec{x}, \vec{y} \in \mathbb{F}^d$ , let*

$$\text{dist}(\vec{x}, \vec{y}) = \frac{|\{i \in [d] : \vec{x}[i] \neq \vec{y}[i]\}|}{d}.$$

*For ease of notation, let  $\text{dist}(\vec{x}) = \text{dist}(\vec{x}, \vec{0}_d)$ .*

Let  $G$  and  $H$  be two groups and  $+$  be the group operator. For any  $\delta \in [0, 1]$  and  $f, g : G \rightarrow H$ , we say  $f$  is  $\delta$ -far from  $g$  if  $\Pr_x[f(x) \neq g(x)] = \delta$ . A function  $f : G \rightarrow H$  is a homomorphism if  $f(x) + f(y) = f(x + y)$  for all  $x, y \in G$ . The follow theorem is from [BLR93, Gol16].

**Theorem 2.2** (Linearity test). *If a function  $f : G \rightarrow H$  satisfies  $\Pr_{x,y}[f(x) + f(y) = f(x + y)] \geq (1 - \delta/2)$  for some small  $\delta$ , then there exists a homomorphism function  $g : G \rightarrow H$  such that  $f$  is at most  $\delta$ -far from  $g$ , i.e.,  $\Pr_x[f(x) = g(x)] \geq (1 - \delta)$ .*

**Definition 2.3** (Constraint Satisfaction Problem (CSP)). *Given an alphabet  $\Gamma$ , an instance of constraint satisfaction problem contains a set of variables  $V = \{v_1, \dots, v_k\}$  and constraints  $\{C_1, C_2, \dots, C_m\}$ . For every  $i \in [m]$ ,  $C_i = (\vec{s}_i, R_i)$ , where  $\vec{s}_i = (v_{j_1}, \dots, v_{j_{\ell_i}})$  is an  $\ell_i$ -tuple of variables for some  $\ell_i \in [k]$  and  $R_i \subseteq \Gamma^{\ell_i}$ . The goal is to find an assignment  $\sigma : V \rightarrow \Gamma$  such that*

- for all  $i \in [m]$ ,  $\sigma(\vec{s}_i) \in R_i$ .

In parameterized complexity, the hypothesis of  $W[1] \neq FPT$  states that no algorithm can, on input a graph  $G$  and a positive integer  $k$ , decide whether  $\omega(G) \geq k$  in  $f(k) \cdot |G|^{O(1)}$  time for any computable function  $f : \mathbb{N} \rightarrow \mathbb{N}$ . A parameterized problem  $L$  is  $W[1]$ -hard if there is a reduction from  $k$ -Clique to this problem such that for every instance  $(G, k)$  of  $k$ -Clique, the reduction outputs an instance  $(x, k')$  of  $L$  in  $f(k) \cdot |G|^{O(1)}$ -time for some computable function  $f : \mathbb{N} \rightarrow \mathbb{N}$  and:

- $(G, k)$  is a yes-instance of  $k$ -Clique if and only if  $(x, k')$  is a yes-instance of  $L$ ,
- $k' \leq g(k)$  for some computable function  $g : \mathbb{N} \rightarrow \mathbb{N}$ .

Obviously, if a parameterized problem  $L$  is  $W[1]$ -hard, then no  $f(k) \cdot |x|^{O(1)}$ -time algorithm can decide whether  $(x, k)$  is a yes-instance of  $L$  unless  $W[1] = FPT$ . We say approximating  $k$ -Clique to a factor of  $c$  is  $W[1]$ -hard if the existence of  $f(k) \cdot |G|^{O(1)}$ -time algorithm that can distinguish  $\omega(G) \geq k$  and  $\omega(G) < k/c$  would imply  $W[1] = FPT$ .

**Definition 2.4** ( $k$ -Vector-Sum). *Given  $k$  sets  $V_1, \dots, V_k$  of vectors and a target vector  $\vec{t}$  in  $\mathbb{F}^m$ , the goal of  $k$ -vector-sum problem is to decide whether there exist  $\vec{v}_1 \in V_1, \dots, \vec{v}_k \in V_k$  such that*

$$\sum_{i \in [k]} \vec{v}_i = \vec{t}.$$

The  $W[1]$ -hardness of  $k$ -Vector-Sum was proved in [ALW13]. For the convenience of the reader, we include a proof in the Appendix.

**Theorem 2.5.**  $k$ -Vector-Sum with  $\mathbb{F} = \mathbb{F}_2$  and  $m = \Theta(k^2 \log n)$  is  $W[1]$ -hard parameterized by  $k$ .

### 3 Gap-reduction from $k$ -Vector-Sum to $k$ -Clique

Given an instance  $(V_1, V_2, \dots, V_k, \vec{t})$  of  $k$ -Vector-Sum over  $\mathbb{F}^m$ . Let  $V = \bigcup_{i \in [k]} V_i$ ,  $n = |V|$  and  $h = \Theta(k^2)$ . By Theorem 2.5, we can assume that  $m = h \log n$ . Let  $\mathbb{F}$  be a finite field with  $|\mathbb{F}| = 4$ . Vectors in the hardness instances from Theorem 2.5 can still be treated as vectors in  $\mathbb{F}^m$ . Since  $V$  only contains vectors in  $\{0, 1\}^m$ , we have

$$\vec{v} \neq c\vec{u} \text{ for any distinct } \vec{v}, \vec{u} \in V \text{ and nonzero } c \in \mathbb{F}. \quad (1)$$

For any  $\ell \in \mathbb{N}$ , select  $\ell$  matrices  $A_1, A_2, \dots, A_\ell \in \mathbb{F}^{h \times m}$  randomly and independently. For every  $\vec{v} \in \mathbb{F}^m$ , let

$$g(\vec{v}) = (A_1 \vec{v}, \dots, A_\ell \vec{v}) \in \mathbb{F}^{h\ell}.$$

For every vector  $\vec{\alpha} \in \mathbb{F}^h$  and  $\vec{v} \in \mathbb{F}^m$ , let

$$f(\vec{\alpha}, \vec{v}) = (\vec{\alpha}^T A_1 \vec{v}, \dots, \vec{\alpha}^T A_\ell \vec{v}) \in \mathbb{F}^\ell.$$

Recall that for any  $d \cdot n$  vectors  $\vec{v} = (v_1, \dots, v_{dn}) \in \mathbb{F}^{dn}$  and  $\vec{a} = (a_1, \dots, a_d) \in \mathbb{F}^d$ ,

$$F(\vec{a}, \vec{v}) = \left( \sum_{i \in [d]} a_i v_i, \dots, \sum_{i \in [d]} a_i v_{i+jd}, \dots, \sum_{i \in [d]} a_i v_{i+(n-1)d} \right) \in \mathbb{F}^n.$$

It follows that  $f(\vec{\alpha}, \vec{v}) = F(\vec{\alpha}, g(\vec{v}))$ . Note that for every  $\vec{v} \in \mathbb{F}^{dn}$ ,  $F(\cdot, \vec{v}) : \mathbb{F}^d \rightarrow \mathbb{F}^n$  is a homomorphism from  $\mathbb{F}^d$  to  $\mathbb{F}^n$ . Every homomorphism  $f$  from  $\mathbb{F}^d$  to  $\mathbb{F}^n$  is also a bitwise linear function, so it can be written as  $f(\cdot) = F(\cdot, \vec{v})$  for some  $\vec{v} \in \mathbb{F}^{dn}$ .

**Lemma 3.1.** *If  $1/10 > (1/|\mathbb{F}|)^{\ell h} \cdot 2^m$ , then with probability at least  $9/10$ , for all nonzero vector  $\vec{v} \in \mathbb{F}^m$ ,  $g(\vec{v}) \neq \vec{0}_{\ell h}$ .*

*Proof.* For any nonzero vector  $\vec{v} \in \mathbb{F}^m$ ,  $\Pr[A_i \vec{v} = \vec{0}_h] = (1/|\mathbb{F}|)^h$ .

$$\Pr[g(\vec{v}) = \vec{0}_{\ell h}] = \prod_{i \in [\ell]} \Pr[A_i \vec{v} = \vec{0}_h] = (1/|\mathbb{F}|)^{\ell h}.$$

With probability at least

$$1 - (1/|\mathbb{F}|)^{\ell h} \cdot 2^m \geq 9/10,$$

$g(\vec{v}) \neq \vec{0}_{\ell h}$  for all nonzero  $\vec{v} \in \mathbb{F}^m$ . □

**Lemma 3.2.** *If  $A \in \mathbb{F}^{h \times m}$  is a random matrix, then for any nonzero vectors  $\vec{b}, \vec{c} \in \mathbb{F}^h$  and distinct  $\vec{v}, \vec{u} \in \mathbb{F}^m$  with  $\vec{v} \neq a\vec{u}$  for any  $a \in \mathbb{F} \setminus \{0\}$ ,*

$$\Pr[\vec{b}^T A \vec{v} = \vec{c}^T A \vec{u}] = 1/|\mathbb{F}|. \quad (2)$$

*Proof.* Let  $B$  be an  $h \times m$  matrix with  $B_{ij} = \vec{b}[i] \cdot \vec{v}[j]$ . Let  $C$  be an  $h \times m$  matrix with  $C_{ij} = \vec{c}[i] \cdot \vec{u}[j]$ . We can treat  $A, B$  as a vector of length  $hm$  and use  $A \cdot B$  denote their dot product. It follows that  $\vec{b}^T A \vec{v} = B \cdot A$  and  $\vec{c}^T A \vec{u} = C \cdot A$ . Since  $\vec{v} \neq a\vec{u}$  for any nonzero  $a \in \mathbb{F}$ , we have  $B - C$  is not a zero matrix. Therefore,

$$\Pr[\vec{b}^T A \vec{v} = \vec{c}^T A \vec{u}] = \Pr[B \cdot A = C \cdot A] = \Pr[(B - C) \cdot A = 0] = 1/|\mathbb{F}|.$$

□

**Lemma 3.3.** *If  $|V|^2 \cdot |\mathbb{F}|^h \cdot (1/|\mathbb{F}|)^\ell < 1/10$ , then with probability at least  $9/10$ ,  $f(\vec{\alpha}, \vec{v}) \neq f(\vec{\alpha}, \vec{u})$  for any distinct  $\vec{v}, \vec{u} \in V$  and nonzero  $\vec{\alpha} \in \mathbb{F}^h$ .*

*Proof.* By (1),  $\vec{v} \neq c\vec{u}$  for any nonzero  $c \in \mathbb{F}$ . Apply Lemma 3.2 with  $\vec{b} = \vec{c} = \vec{\alpha}$ , we get

$$\Pr[f(\vec{\alpha}, \vec{v}) = f(\vec{\alpha}, \vec{u})] = \prod_{i \in [\ell]} \Pr[\vec{\alpha}^T A_i \vec{v} = \vec{\alpha}^T A_i \vec{u}] = (1/|\mathbb{F}|)^\ell.$$

There are at most  $|V|^2$  pairs of  $(\vec{v}, \vec{u})$  and at most  $|\mathbb{F}|^d$  choices of  $\vec{\alpha}$ . Since  $|V|^2 \cdot |\mathbb{F}|^d \cdot (1/|\mathbb{F}|)^\ell < 1/10$ , with probability at least  $9/10$ ,  $f(\vec{\alpha}, \vec{v}) \neq f(\vec{\alpha}, \vec{u})$  for all nonzero  $\vec{\alpha} \in \mathbb{F}^h$  and distinct  $\vec{v}, \vec{u} \in V$ .  $\square$

**Lemma 3.4.** *If  $|V|^3 \cdot |\mathbb{F}|^{2h} \cdot (1/|\mathbb{F}|)^\ell < 1/10$ , then with probability at least  $9/10$ ,  $f(\vec{\alpha}, \vec{v}) + f(\vec{\alpha}', \vec{u}) \neq f(\vec{\alpha} + \vec{\alpha}', \vec{w})$  for any distinct  $\vec{v}, \vec{u}, \vec{w} \in V$  and nonzero  $\vec{\alpha}, \vec{\alpha}'$ .*

*Proof.* Observe that  $\Pr[f(\vec{\alpha}, \vec{v}) + f(\vec{\alpha}', \vec{u}) = f(\vec{\alpha} + \vec{\alpha}', \vec{w})] = \Pr[f(\vec{\alpha}, \vec{v} - \vec{w}) = f(\vec{\alpha}', \vec{w} - \vec{u})]$  and  $\vec{w} - \vec{u} = \vec{w} + \vec{u} \neq \vec{v} + \vec{w} = \vec{v} - \vec{w}$ . Since  $\vec{w} - \vec{u}$  and  $\vec{v} - \vec{w}$  are vectors in  $\{0, 1\}^m \subseteq \mathbb{F}^m$ ,  $\vec{w} - \vec{u} \neq (\vec{v} - \vec{w})$  implies  $\vec{w} - \vec{u} \neq a(\vec{v} - \vec{w})$  for any  $a \in \mathbb{F} \setminus \{0\}$ . By Lemma 3.2,  $\Pr[f(\vec{\alpha}, \vec{v} - \vec{w}) = f(\vec{\alpha}', \vec{w} - \vec{u})] \leq (1/|\mathbb{F}|)^\ell$ . There are at most  $|V|^3$  pairs of  $(\vec{v}, \vec{u}, \vec{w})$  and at most  $|\mathbb{F}|^{2d}$  choices of  $\vec{\alpha}, \vec{\alpha}'$ . Since  $|V|^3 \cdot |\mathbb{F}|^{2d} \cdot (1/|\mathbb{F}|)^\ell < 1/10$ , with probability at least  $9/10$ ,  $f(\vec{\alpha}, \vec{v}) + f(\vec{\alpha}', \vec{u}) \neq f(\vec{\alpha} + \vec{\alpha}', \vec{w})$  for any distinct  $\vec{v}, \vec{u}, \vec{w} \in V$  and nonzero  $\vec{\alpha}, \vec{\alpha}'$ .  $\square$

*Construction of the CSP.* Let  $\ell = 2 \log n + 2h$ . Then for large  $n$ ,

$$(1/|\mathbb{F}|)^{\ell h} \cdot 2^{h \log n} = 4^{-2h \log n - 2h} \cdot 2^{h \log n} \leq 2^{-3h \log n} < 1/10,$$

and

$$|V|^3 \cdot |\mathbb{F}|^{2h} \cdot (1/|\mathbb{F}|)^\ell = n^3 \cdot 4^{2h} \cdot 4^{-2 \log n - 2h} \leq 1/n \leq 1/10.$$

By Lemma 3.3, Lemma 3.4 and Lemma 3.1, with probability at least  $7/10$ ,  $g(\vec{v}) \neq \vec{0}_{\ell h}$  for all  $\vec{v} \in \mathbb{F}^{h \log n}$ ,  $f(\vec{\alpha}, \vec{v}) \neq f(\vec{\alpha}, \vec{u})$  and  $f(\vec{\alpha}, \vec{v}) + f(\vec{\alpha}', \vec{u}) \neq f(\vec{\alpha} + \vec{\alpha}', \vec{w})$  for all distinct  $\vec{v}, \vec{u}, \vec{w} \in V$  and nonzero  $\vec{\alpha}, \vec{\alpha}'$ .

Construct a CSP instance  $I$  with  $|\mathbb{F}|^{kh}$  variables  $\{x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} : \vec{\alpha}_1 \dots \vec{\alpha}_k \in \mathbb{F}^h\}$ . The alphabet of this CSP is  $\Gamma = \mathbb{F}^\ell$ . If the instance is a yes-instance, then each  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$  is expected to take the value  $f(\vec{\alpha}_1, \vec{v}_1) + \dots + f(\vec{\alpha}_k, \vec{v}_k)$  for some a solution  $\vec{v}_1, \dots, \vec{v}_k$  to the  $k$ -Vector-Sum problem. We now describe three types of constraints.

**(C1)** For all  $\vec{\alpha}_1, \dots, \vec{\alpha}_k$  and  $\vec{\beta}_1, \dots, \vec{\beta}_k$ , check if  $x_{\vec{\alpha}_1 + \vec{\beta}_1, \dots, \vec{\alpha}_k + \vec{\beta}_k} = x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} + x_{\vec{\beta}_1, \dots, \vec{\beta}_k}$ . In other words, the relation for variable tuple  $(x_{\vec{\alpha}_1 + \vec{\beta}_1, \dots, \vec{\alpha}_k + \vec{\beta}_k}, x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}, x_{\vec{\beta}_1, \dots, \vec{\beta}_k})$  is

$$R_{x_{\vec{\alpha}_1 + \vec{\beta}_1, \dots, \vec{\alpha}_k + \vec{\beta}_k}, x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}, x_{\vec{\beta}_1, \dots, \vec{\beta}_k}} = \{(\vec{a}, \vec{b}, \vec{c}) \in \Gamma^3 : \vec{a} = \vec{b} + \vec{c}\}.$$

**(C2)** For every  $i \in [k]$  and  $\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\alpha} \in \mathbb{F}^h$ , check if  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_i + \vec{\alpha}, \dots, \vec{\alpha}_k} - x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = f(\vec{\alpha}, \vec{v})$  for some  $\vec{v} \in V_i$ . In other words, the constraint between  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_i + \vec{\alpha}, \dots, \vec{\alpha}_k}$  and  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$  is

$$R_{x_{\vec{\alpha}_1, \dots, \vec{\alpha}_i + \vec{\alpha}, \dots, \vec{\alpha}_k}, x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}} = \{(\vec{a}, \vec{b}) \in \Gamma^2 : \vec{a} - \vec{b} = f(\vec{\alpha}, \vec{v}) \text{ for some } \vec{v} \in V_i\}.$$

**(C3)** For all  $\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\alpha} \in \mathbb{F}^h$ , check if  $x_{\vec{\alpha}_1 + \vec{\alpha}, \dots, \vec{\alpha}_k + \vec{\alpha}} - x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = f(\vec{\alpha}, \vec{t})$ . That is,

$$R_{x_{\vec{\alpha}_1 + \vec{\alpha}, \dots, \vec{\alpha}_k + \vec{\alpha}}, x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}} = \{(\vec{a}, \vec{b}) \in \Gamma^2 : \vec{a} - \vec{b} = f(\vec{\alpha}, \vec{t})\}.$$

Constraints of the form  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_i + \vec{\alpha}, \dots, \vec{\alpha}_k} - x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = f(\vec{\alpha}, \vec{v})$  in (C2) are also called (C2)- $i$ -type or (C2)- $i$ - $\vec{\alpha}$ -type constraints. Similarly, constraints of the form  $x_{\vec{\alpha}_1 + \vec{\alpha}, \dots, \vec{\alpha}_k + \vec{\alpha}} - x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = f(\vec{\alpha}, \vec{t})$  are called (C3)- $\vec{\alpha}$ -type constraints.

**Lemma 3.5.** *If the  $k$ -Vector-Sum instance has a solution, then so does  $I$ . If the  $k$ -Vector-Sum instance has no solution, then there exists a constant  $\epsilon > 0$  such that for every assignment to the variables of  $I$  one of the followings must hold:*

- $\epsilon/2$ -fraction of constraints of (C1) are not satisfied,
- there exists  $i \in [k]$  such that  $\epsilon^2$ -fraction of (C2)- $i$ -type constraints are not satisfied.
- $\epsilon$ -fraction of constraints of (C3) are not satisfied.

*Proof.* If the  $k$ -Vector-Sum instance has a solution  $\vec{v}_1 \in V_1, \dots, \vec{v}_k \in V_k$ , then let  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = f(\vec{\alpha}_1, \vec{v}_1) + \dots + f(\vec{\alpha}_k, \vec{v}_k)$ . It is easy to check that all the constraints are satisfied.

Now suppose that the  $k$ -Vector-Sum instance has no solution. Fix any assignment  $\vec{x} \in \mathbb{F}^\ell$ . If  $\epsilon/2$ -fraction of constraints in (C1) are not satisfied, then we are done. Otherwise  $(1 - \epsilon/2)$ -fraction of (C1) constraints are satisfied. By the linearity test [BLR93] and (C1), there exist  $\vec{c}_1, \dots, \vec{c}_k \in \mathbb{F}^{h\ell}$  such that for  $(1 - \epsilon)|\mathbb{F}^{kh}|$  many choices of  $(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in \mathbb{F}^{kh}$ ,  $\vec{x}_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = F(\vec{\alpha}_1, \vec{c}_1) + F(\vec{\alpha}_2, \vec{c}_2) + \dots + F(\vec{\alpha}_k, \vec{c}_k)$ . Let

$$A = \{(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in \mathbb{F}^{kh} : \vec{x}_{\vec{\alpha}_1, \dots, \vec{\alpha}_k} = F(\vec{\alpha}_1, \vec{c}_1) + F(\vec{\alpha}_2, \vec{c}_2) + \dots + F(\vec{\alpha}_k, \vec{c}_k)\}.$$

We have that  $|A| \geq (1 - \epsilon)|\mathbb{F}^{kh}|$ .

Obviously, there are two cases:

- Either for every  $i \in [k]$ , there exists  $\vec{v}_i \in V_i$  such that  $\vec{c}_i = g(\vec{v}_i)$ ,
- or there exists  $i \in [k]$  such that  $\vec{c}_i \neq g(\vec{v})$  for all  $\vec{v} \in V_i$ .

In the later case, we will show that at least an  $\epsilon^2$ -fraction of (C2)- $i$ -type constraints are not satisfied. Call a vector  $\vec{\alpha} \in \mathbb{F}^h$  good if at least  $(1 - \epsilon)$ -fraction of (C2)- $i$ - $\vec{\alpha}$ -type constraints are satisfied. For every nonzero vector  $\vec{\alpha} \in \mathbb{F}^h$ ,  $\mathbb{F}^{kh}$  can be partitioned into two disjoint sets  $X_{\vec{\alpha}}^-$  and  $X_{\vec{\alpha}}^+$  such that  $X_{\vec{\alpha}}^+ = \{(\vec{a}_1, \dots, \vec{a}_i + \vec{\alpha}, \dots, \vec{a}_k) : (\vec{a}_1, \dots, \vec{a}_k) \in X_{\vec{\alpha}}^-\}$ . Now suppose  $\vec{\alpha}$  is a good vector. There are  $|\mathbb{F}^{kh}|/2$  constraints of (C2)- $i$ - $\vec{\alpha}$ -type. We construct a bipartite graph on  $X_{\vec{\alpha}}^-$  and  $X_{\vec{\alpha}}^+$ . Two vertices  $(\vec{a}_1, \dots, \vec{a}_k) \in X_{\vec{\alpha}}^-$  and  $(\vec{a}_1, \dots, \vec{a}_i + \vec{\alpha}, \dots, \vec{a}_k) \in X_{\vec{\alpha}}^+$  are adjacent if  $(\vec{x}_{\vec{a}_1, \dots, \vec{a}_k}, \vec{x}_{\vec{a}_1, \dots, \vec{a}_i + \vec{\alpha}, \dots, \vec{a}_k})$  satisfies the constraint between them. Since  $\vec{\alpha}$  is a good vector, there are  $(1 - \epsilon)|\mathbb{F}^{kh}|/2$  edges between  $X_{\vec{\alpha}}^-$  and  $X_{\vec{\alpha}}^+$ . These edges form a matching  $M$ . Observe that  $\min\{|A \cap X_1|, |A \cap X_2|\} \geq (1/2 - \epsilon)|\mathbb{F}^{kh}|$ . Since the size of matching  $M$  is  $(1 - \epsilon)|\mathbb{F}^{kh}|/2$ , when  $6\epsilon < 1$ ,  $M$  contains an edge whose endpoints are both in  $A$ . In other words, there exists  $(\vec{a}_1, \dots, \vec{a}_k) \in A$  such that  $(\vec{a}_1, \dots, \vec{a}_i + \vec{\alpha}, \dots, \vec{a}_k) \in A$  and  $(\vec{x}_{\vec{a}_1, \dots, \vec{a}_i + \vec{\alpha}, \dots, \vec{a}_k}, \vec{x}_{\vec{a}_1, \dots, \vec{a}_k})$  satisfies the (C2) constraint. By (C2) and Lemma 3.3, we deduce that  $F(\vec{\alpha}, \vec{c}_i) = f(\vec{\alpha}, \vec{v})$  for some unique  $\vec{v} \in V_i$ . To summarize, for every good vector  $\vec{\alpha}$ , there exists a vector  $\vec{v}_{\vec{\alpha}} \in V_i$  such that  $F(\vec{\alpha}, \vec{c}_i) = f(\vec{\alpha}, \vec{v}_{\vec{\alpha}})$ .

Next, we show that there are at most  $(1 - \epsilon)|\mathbb{F}^h|$  good vectors, and hence at most  $(1 - \epsilon) + \epsilon(1 - \epsilon) = (1 - \epsilon^2)$ -fraction of constraints of (C2)- $i$ -type are satisfied. Otherwise, pick an arbitrary good vector  $\vec{\alpha}$ . All the vectors in  $\mathbb{F}^h - \{0, \vec{\alpha}\}$  can be partitioned into two sets  $X_1$  and  $X_2$  such that  $X_1 = X_2 + \vec{\alpha}$ . There exists  $Y_1 \subseteq X_1$  such that  $Y_1$  is a set of good vectors and  $|Y_1| \geq (1/2 - \epsilon)|\mathbb{F}^h|$ . Since  $\vec{c}_i \neq g(\vec{v}_{\vec{\alpha}})$ , there is a set  $X$  of size at most  $|\mathbb{F}|^{h-1}$  such that for all  $\vec{x} \in \mathbb{F}^h \setminus X$ ,  $F(\vec{x}, \vec{c}_i) \neq F(\vec{x}, g(\vec{v}_{\vec{\alpha}}))$ . Since  $(1 - \epsilon)|\mathbb{F}^h|$  vectors are good, there exists a set  $Z \subseteq Y_2 = Y_1 + \vec{\alpha}$  such that  $|Z| \leq \epsilon|\mathbb{F}^h|$  and all the bad vectors of  $Y_2$  are in  $Z$ . When  $2\epsilon + 1/|\mathbb{F}| < 1/2$ , we have  $|Y_1| - |X| - |Z| > 0$ . Thus there exists a vector  $\vec{\alpha}' \in Y_1 - X - (Z + \vec{\alpha})$ . According to the definitions,  $\vec{\alpha}'$  and  $\vec{\alpha} + \vec{\alpha}'$  are good and  $\vec{v}_{\vec{\alpha}} \neq \vec{v}_{\vec{\alpha}'}$ . Since  $\vec{\alpha} + \vec{\alpha}'$  is good, there exists  $\vec{u} \in V_i$  such that  $F(\vec{\alpha}' + \vec{\alpha}, \vec{c}_i) = f(\vec{\alpha}' + \vec{\alpha}, \vec{u})$ . Note that  $\vec{u} \neq \vec{v}_{\vec{\alpha}}$ , otherwise by  $F(\vec{\alpha}, \vec{c}_i) = f(\vec{\alpha}, \vec{v}_{\vec{\alpha}})$ , we can deduce that  $F(\vec{\alpha}', \vec{c}_i) = f(\vec{\alpha}', \vec{u})$ , which implies  $\vec{u} = \vec{v}_{\vec{\alpha}'} \neq \vec{v}_{\vec{\alpha}}$ , that is impossible. So we have  $f(\vec{\alpha}' + \vec{\alpha}, \vec{u}) = F(\vec{\alpha}, \vec{c}_i) + F(\vec{\alpha}', \vec{c}_i) = f(\vec{\alpha}, \vec{v}_{\vec{\alpha}}) + f(\vec{\alpha}', \vec{v}_{\vec{\alpha}'})$ , where  $\vec{\alpha}, \vec{\alpha}'$  are nonzero vectors and  $\vec{v}_{\vec{\alpha}}, \vec{v}_{\vec{\alpha}'}, \vec{u}$  are distinct, contradicting Lemma 3.4.

Now assume that  $\vec{c}_i = g(\vec{v}_i)$  for every  $i \in [k]$ . Note that  $\vec{v}_1 + \dots + \vec{v}_k \neq \vec{t}$ . By Lemma 3.1,  $g(\sum_{i \in [k]} \vec{v}_i) \neq g(\vec{t})$ . There exists a set  $B \subseteq \mathbb{F}^h$  such that  $|B| \geq (1 - 1/|\mathbb{F}|) \cdot |\mathbb{F}^h|$  and for all  $\vec{\alpha} \in B$ ,

$$\sum_{i \in [k]} F(\vec{\alpha}, \vec{c}_i) = \sum_{i \in [k]} F(\vec{\alpha}, g(\vec{v}_i)) = F(\vec{\alpha}, \sum_{i \in [k]} g(\vec{v}_i)) \neq F(\vec{\alpha}, g(\vec{t})) = f(\vec{\alpha}, \vec{t}).$$

Notice that  $|A| \geq (1 - \epsilon)|\mathbb{F}^{kh}|$ . We have

$$|\{(\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\alpha}) : (\vec{\alpha}_1, \dots, \vec{\alpha}_k), (\vec{\alpha}_1 + \vec{\alpha}, \dots, \vec{\alpha}_k + \vec{\alpha}) \in A, \vec{\alpha} \in B\}| \geq (1 - 1/|\mathbb{F}|) \cdot (1 - 2\epsilon) \cdot |\mathbb{F}|^{k(h+1)}.$$

This implies that at least  $(1 - 1/|\mathbb{F}|) \cdot (1 - \epsilon)|\mathbb{F}|^{k(h+1)} > \epsilon|\mathbb{F}|^{k(h+1)}$  constraints in (C3) are not satisfied.  $\square$

*Construction of the Gap-clique instance.* For every  $\vec{\alpha}_1, \dots, \vec{\alpha}_k \in \mathbb{F}^h$  and  $\vec{\beta}_1, \dots, \vec{\beta}_k \in \mathbb{F}^h$ , introduce a vertex set  $V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\beta}_1, \dots, \vec{\beta}_k} = \{(\vec{x}, \vec{y}, \vec{z}) : \vec{x}, \vec{y}, \vec{z} \in \mathbb{F}^\ell, \vec{x} = \vec{y} + \vec{z}\}$ . Each vertex in this set is corresponding to an assignment to three variables  $x_{\vec{\alpha}_1 + \vec{\beta}_1, \dots, \vec{\alpha}_k + \vec{\beta}_k}$ ,  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$  and  $x_{\vec{\beta}_1, \dots, \vec{\beta}_k}$  which satisfies the constraint of (C1). For every variables  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$  and  $i \in [|\mathbb{F}|^{kh}]$ , introduce a vertex set  $V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i} = \mathbb{F}^\ell$ . Each vertex in  $V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i}$  is an assignment to the variable  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$ .

Construct a graph  $G'$  on vertex  $(\bigcup_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\beta}_1, \dots, \vec{\beta}_k} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\beta}_1, \dots, \vec{\beta}_k}) \cup (\bigcup_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i})$ . Two vertices in  $G'$  are adjacent unless they are corresponding to inconsistent assignments or they do not satisfy the constraint between the variables they assigned to.

**Lemma 3.6.** *If the  $k$ -Vector-Sum instance has a solution, then the graph  $G'$  contains a clique of size  $2|\mathbb{F}|^{2kh}$ .*

*Proof.* For every vertex set, select the vertex corresponding to the assignment. According to the definition, these vertices form a clique of size  $2|\mathbb{F}|^{2kh}$ .  $\square$

**Lemma 3.7.** *If the  $k$ -Vector-Sum instance has no solution, then the graph  $G'$  contains no clique of size  $(1 - \epsilon')2|\mathbb{F}|^{2kh}$  for some small constant  $\epsilon' > 0$ .*

*Proof.* Let  $\epsilon$  be the constant in Lemma 3.5. Pick a small  $\epsilon'$  such that  $4\epsilon' < \min\{\epsilon/2, \epsilon^2\}$ . Let  $X$  be the clique in the graph of size larger than  $(1 - \epsilon')2|\mathbb{F}|^{2kh}$ . We have

$$|X \cap \bigcup_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\beta}_1, \dots, \vec{\beta}_k \in \mathbb{F}^h} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, \vec{\beta}_1, \dots, \vec{\beta}_k}| \geq (1 - 2\epsilon')|\mathbb{F}|^{2kh}. \quad (3)$$

In addition, since  $|X| > (1 - \epsilon')2|\mathbb{F}|^{2kh}$ , there exists an index  $i \in [|\mathbb{F}|^{kh}]$  such that  $X$  contains more than  $(1 - 2\epsilon')|\mathbb{F}|^{kh}$  vertices with respect to index  $i$ . We will prove that this is impossible using the following Claim 1.

*Claim 1.* For every  $i \in [|\mathbb{F}|^{kh}]$ ,  $|X \cap \bigcup_{\vec{\alpha}_1, \dots, \vec{\alpha}_k \in \mathbb{F}^h} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i}| \leq (1 - 2\epsilon')|\mathbb{F}|^{kh}$ .

*Proof of Claim 1.* Fix an index  $i \in [|\mathbb{F}|^{kh}]$ . Define an assignment  $\sigma_X$  as follows. For every  $\vec{\alpha}_1, \dots, \vec{\alpha}_k \in \mathbb{F}^h$ ,  $\sigma_X(x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}) = v$  if  $\{v\} = X \cap V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i}$ , otherwise  $\sigma_X(x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}) = \vec{0}_\ell$ . By (3) and the definition of edge set,  $\delta_X$  satisfies  $(1 - 2\epsilon')$ -fraction of constraints in (C1). By Lemma 3.5, either  $\sigma_X$  satisfies at most  $(1 - 4\epsilon')$ -fraction of constraints in (C3) or there exists an  $j \in [k]$  such that  $\sigma_X$  satisfies at most  $(1 - 4\epsilon')$ -fraction of constraints of (C2)- $j$  type.

- Assume that  $\sigma_X$  satisfies at most  $(1 - 4\epsilon')$ -fraction of constraints in (C3). We say a vector  $\vec{\alpha} \in \mathbb{F}^h$  is  $\delta$ -good if more than  $\delta$ -fraction of (C3)- $\vec{\alpha}$ -type constraints are satisfied by  $\sigma_X$ . There exists a vector  $\vec{\alpha} \in \mathbb{F}^h$  that is not  $(1 - 4\epsilon')$ -good, otherwise  $\delta_X$  satisfies more than  $(1 - 4\epsilon')$ -fraction of constraints in (C3), contradicting our assumption. Now consider a partition  $\mathbb{F}^{kh} = V_{\vec{\alpha}}^- \cup V_{\vec{\alpha}}^+$  with  $V_{\vec{\alpha}}^+ = V_{\vec{\alpha}}^- + (\vec{\alpha}, \dots, \vec{\alpha})$ . Since  $\vec{\alpha}$  is not  $(1 - 4\epsilon')$ -good, there are at most  $(1 - 4\epsilon')|\mathbb{F}|^{kh}/2$  tuples  $(\vec{\alpha}_1, \dots, \vec{\alpha}_k)$  in  $V_{\vec{\alpha}}^-$  such that  $\delta_X$  satisfies the (C3) constraint between  $x_{\vec{\alpha}_1, \dots, \vec{\alpha}_k}$  and  $x_{\vec{\alpha}_1 + \vec{\alpha}, \dots, \vec{\alpha}_k + \vec{\alpha}}$ . Let

$$X_{\vec{\alpha}}^- = X \cap \bigcup_{(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in V_{\vec{\alpha}}^-} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i} \text{ and } X_{\vec{\alpha}}^+ = X \cap \bigcup_{(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in V_{\vec{\alpha}}^+} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i}$$

It follows that  $\min\{|X_{\vec{\alpha}}^-|, |X_{\vec{\alpha}}^+|\} \leq (1 - 4\epsilon')|\mathbb{F}|^h/2$ . Thus  $|X| \leq (1/2 + (1 - 4\epsilon')/2)|\mathbb{F}|^{kh} = (1 - 2\epsilon')|\mathbb{F}|^{kh}$ .

- Now assume that  $\sigma_X$  satisfies at most  $(1 - 4\epsilon')$ -fraction of constraints of (C2)- $j$  type for some  $j \in [k]$ . Similarly, for every  $\vec{\alpha} \in \mathbb{F}^h$ , we say  $\vec{\alpha}$  is  $\delta$ -good if more than  $\delta$ -fraction of (C2)- $j$ - $\vec{\alpha}$ -type constraints are satisfied by  $\sigma_X$ . There exists  $\vec{\alpha} \in \mathbb{F}^h$  that is not  $(1 - 4\epsilon')$ -good,



otherwise  $\delta_X$  satisfies more than  $(1 - 4\epsilon')$ -fraction of constraints of type (C2)- $j$ , contradicting our assumption. Now consider a partition  $\mathbb{F}^{kh} = V_{\vec{\alpha}}^- \cup V_{\vec{\alpha}}^+$  with

$$V_{\vec{\alpha}}^+ = \{(\vec{\alpha}_1, \dots, \vec{\alpha}_i + \vec{\alpha}, \dots, \vec{\alpha}_k) : (\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in V_{\vec{\alpha}}^-\}.$$

Let

$$X_{\vec{\alpha}}^- = X \cap \bigcup_{(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in V_{\vec{\alpha}}^-} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i} \text{ and } X_{\vec{\alpha}}^+ = X \cap \bigcup_{(\vec{\alpha}_1, \dots, \vec{\alpha}_k) \in V_{\vec{\alpha}}^+} V_{\vec{\alpha}_1, \dots, \vec{\alpha}_k, i}.$$

Since  $\vec{\alpha}$  is not  $(1 - 4\epsilon')$ -good, we have  $\min\{|X_{\vec{\alpha}}^-|, |X_{\vec{\alpha}}^+|\} \leq (1 - 4\epsilon')|\mathbb{F}|^h/2$ . Thus  $|X| \leq (1/2 + (1 - 4\epsilon')/2)|\mathbb{F}^{kh}| = (1 - 2\epsilon')|\mathbb{F}|^{kh}$ .

□

### 3.1 Putting all together

For any instance  $(G, k)$  of  $k$ -Clique, we use Theorem 2.5 to reduce it to an instance  $(k', V_1, \dots, V_{k'}, \vec{t})$  of  $k'$ -Vector-Sum with  $k' = \Theta(k^2)$ . Then we use the reduction describe to obtain a graph  $G'$  with  $h = k'^2 = \Theta(k^4)$  and small  $\epsilon > 0$  in  $2^{k^{O(1)}} \cdot |G|^{O(1)}$ -time. By Lemma 3.6 and Lemma 3.7, we have

- if  $\omega(G) \geq k$ , then  $\omega(G') \geq 2^{4kh+1}$ ,
- if  $\omega(G) < k$ , then with probability at least  $7/10$ ,  $\omega(G') < (1 - \epsilon)2^{4kh+1}$ .

Using the graph product method, we can amplify the gap to any constant.

### 3.2 Derandomization

To derandomize the reduction, we need to construct  $O(\log n + h)$  matrices  $A_1, \dots, A_\ell \in \mathbb{F}^{h \times m}$  such that the following conditions are satisfied. For all nonzero  $\vec{v} \in \mathbb{F}^m = \mathbb{F}^{h \log n}$ ,

$$(A_1 \vec{v}, \dots, A_\ell \vec{v}) \neq \vec{0}_{\ell h} \quad (4)$$

For all distinct  $\vec{v}, \vec{u} \in V$  and nonzero  $\vec{\alpha} \in \mathbb{F}^h$ ,

$$(\vec{\alpha}^T A_1 \vec{v}, \dots, \vec{\alpha}^T A_\ell \vec{v}) \neq (\vec{\alpha}^T A_1 \vec{u}, \dots, \vec{\alpha}^T A_\ell \vec{u}) \quad (5)$$

For all distinct  $\vec{v}, \vec{u}, \vec{w} \in V$  and nonzero  $\vec{\alpha}, \vec{\alpha}' \in \mathbb{F}^h$

$$(\vec{\alpha}^T A_1(\vec{v} + \vec{w}), \dots, \vec{\alpha}^T A_\ell(\vec{v} + \vec{w})) \neq (\vec{\alpha}'^T A_1(\vec{u} + \vec{w}), \dots, \vec{\alpha}'^T A_\ell(\vec{u} + \vec{w})) \quad (6)$$

Let  $A_i \in \mathbb{F}^{h \times m}$  be the matrix such that  $A_i \vec{v}$  is the projection of  $\vec{v}$  onto the its subvector with coordinates between  $1 + (i - 1)h$  and  $ih$ . Then  $A_1, \dots, A_{\log n}$  satisfy (4). It remains to construct another  $O(\log n + h)$  matrices  $A'_1, \dots, A'_\ell \in \mathbb{F}^{h \times m}$  satisfying (5) and (6). Then their union  $A_1, \dots, A_{\log n}, A'_1, \dots, A'_\ell$  would satisfy all the conditions. Note that we can think of a matrix in  $\mathbb{F}^{h \times m}$  as an  $hm$ -dimension vector. The task can be formulated as given  $N = |\mathbb{F}|^{2h} \cdot n^{O(1)}$  vectors  $C_1, \dots, C_N \in \mathbb{F}^{hm}$ , find  $O(\log n + h)$  vectors  $A'_1, \dots, A'_\ell \in \mathbb{F}^{hm}$  such that for every  $i \in [N]$ , there exists  $A'_j$  such that  $A'_j \cdot C_i \neq 0$ .

We show that, in  $N^{O(1)}$ -time, a vector  $A \in \mathbb{F}^{hm}$  can be found such that there are at most  $N/|\mathbb{F}|$  indices  $i \in [N]$  satisfying  $A \cdot C_i = 0$ . Then we apply this algorithm  $\log N / \log |\mathbb{F}|$  times to obtain the vectors  $A'_1, \dots, A'_\ell$ . The vector  $A$  can be found using the method of conditional probabilities [Juk11, AS04]. Let  $A$  be a vector with  $A[i]$  selected randomly and independently from  $\mathbb{F}$ . Define a random variable  $X = |\{i \in [N] : A \cdot C_i = 0\}|$ . We have  $E[X] = N/|\mathbb{F}|$ . For  $a_1, \dots, a_i \in \mathbb{F}$ , let  $X[a_1, \dots, a_i] = |\{i \in [N] : A \cdot C_i = 0, A[1] = a_1, \dots, A[i] = a_i\}|$ . We have

$$E[X|a_1, \dots, a_i] = \sum_{x \in \mathbb{F}} E[X|a_1, \dots, a_i, x]/|\mathbb{F}| \geq \min\{E[X|a_1, \dots, a_i, x] : x \in \mathbb{F}\}.$$

For each  $i \in [hm]$ ,  $E[X|a_1, \dots, a_i]$  can be computed in  $N^{O(1)}$ -time. For each  $i \in [hm]$ , we pick the value  $a_i$  to minimize  $E[X|a_1, \dots, a_i]$ . We have  $E[X|a_1, \dots, a_{hm}] \leq N/|\mathbb{F}|$  and the vector  $A$  with  $A[i] = a_i (i \in [hm])$  is our target vector.

## 4 Conclusion

This paper constructs a PCP verifier which always accepts yes-instances and with probability  $\Theta(1/k)$  rejects no-instances for a  $W[1]$ -hard problem and shows how to create a constant gap for  $\omega(G)$  using this PCP. I hope that the technique of this paper will help obtain a parameterized version of PCP theorem, e.g. the Parameterized Inapproximability Hypothesis (PIH) [LRSZ20].

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## Appendix

**Theorem 4.1** (Theorem 2.5 restated).  *$k$ -Vector-Sum with  $\mathbb{F} = \mathbb{F}_2$  and  $m = \Theta(k^2 \log n)$  is  $W[1]$ -hard parameterized by  $k$ .*

*Proof.* We construct a reduction from  $k$ -Multi-Color-Clique to  $(k + \binom{k}{2})$ -Vector-Sum. Let  $(G, k)$  be an instance of  $k$ -Multi-Color-Clique with  $V(G) = V_1 \cup V_2 \cup \dots \cup V_k$ . Set  $n = |V(G)| + 1$ . For every  $v \in V(G)$ , let  $\sigma(v) \in \{0, 1\}^{\log n}$  be the binary encoding of  $v$ . Since  $n > |V(G)|$ , we can assume that  $\sigma(v) \neq \vec{0}_{\log n}$  for all  $v \in V(G)$ . Let  $f : \binom{[k]}{2} \rightarrow \binom{[k]}{2}$  be a bijection. For every  $i \in [k]$  and  $v \in V(G)$ , let

$$\vec{\eta}_{v,i} = (\underbrace{0, \dots, 0, \sigma(v), 0, \dots, 0}_{k \log n}) \text{ and } \vec{\gamma}_{v,i} = (\underbrace{\sigma(v), \dots, \sigma(v), \vec{0}_{\log n}, \sigma(v), \dots, \sigma(v)}_{k \log n})$$

For any distinct  $i, j \in [k]$ , let

$$\vec{\theta}_{i,j} = (\underbrace{0, \dots, 0, 1, 0, \dots, 0}_{k(k-1)/2}) \text{ and } \vec{\delta}_i = (\underbrace{0, \dots, 0, 1, 0, \dots, 0}_k).$$

For every edge  $e = \{v, u\}$  with  $v \in V_i$  and  $u \in V_j$ , let

$$\vec{w}_e = \vec{0}_k \circ \vec{\theta}_{i,j} \circ (\underbrace{\underbrace{0, \dots, 0}_{(i-1)k \log n}, \vec{\eta}_{v,j}, 0, \dots, 0, \vec{\eta}_{u,i}, 0, \dots, 0}_{(j-1)k \log n}}_{k^2 \log n}).$$

For every  $v \in V_i$ , let

$$\vec{w}_v = \vec{\delta}_i \circ \vec{0}_{k(k-1)/2} \circ (\underbrace{0, \dots, 0}_{(i-1)k \log n}, \vec{\gamma}_{v,i}, 0, \dots, 0)_{k^2 \log n}$$

The instance of vector sum is defined as follows.

- The target vector is  $\vec{t} = \vec{1}_{k+k(k-1)/2} \circ \vec{0}_{k^2 \log n}$ .
- There are  $k(k-1)/2 + k$  sets of vectors.

– For every  $\{i, j\} \in \binom{[k]}{2}$ , let

$$W_{ij} = \{\vec{w}_e : e = \{v, u\} \text{ is an edge in } G \text{ with } v \in V_i \text{ and } u \in V_j\}.$$

– For every  $i \in [k]$ , let

$$W_i = \{\vec{w}_v : v \in V_i\}.$$

If  $(G, k)$  is a yes-instance, then there exist  $v_1 \in V_1, \dots, v_k \in V_k$  such that  $\{v_1, \dots, v_k\}$  induces a  $k$ -clique in  $G$ . It is easy to check that the sum of  $\vec{w}_{v_i v_j}$ 's and  $\vec{w}_{v_i}$ 's is equal to  $\vec{t}$ .

On the other hand, if there exist  $\vec{w}_{ij} \in W_{ij}$  and  $\vec{w}_i \in W_i$  such that

$$\sum \vec{w}_{ij} + \sum \vec{w}_i = \vec{t}.$$

Each  $w_i$  is corresponding to a vertex  $v_i \in V_i$ . Each  $w_{ij}$  is corresponding to an edge  $e_{ij}$  between  $V_i$  and  $V_j$ . It is easy to see that  $v_i$  is an endpoint of  $e_{ij}$  for all  $j \in [k] \setminus \{i\}$ . Therefore they form a clique of size  $k$ .  $\square$