# One-key Double-Sum MAC with Beyond-Birthday Security 

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#### Abstract

MACs (Message Authentication Codes) are widely adopted in communication systems to ensure data integrity and data origin authentication, e.g. CBC-MACs in the ISO standard 9797-1. However, all the current designs either suffer from birthday attacks or require long key sizes. In this paper, we focus on designing beyond-birthday-bound MAC modes with a single key, and investigate their design principles. First, we review the current proposals, e.g. 3kf9 and PMAC_Plus, and identify that the security primarily comes from the construction of a cover-free function and the advantage of the sum of PRPs. The main challenge in reducing their key size is to find a mechanism to carefully separate the block cipher inputs to the cover-free construction and the sum of PRPs that work in cascade with such a construction. Secondly, we develop several tools on sampling distributions that are quite useful in analysis of the MAC mode of operations and by which we unify the proofs for three/two-key beyond-birthday-bound MACs. Thirdly, we establish our main theorem that upper-bounds the PRF security of the one-key constructions by extended-cover-free, pseudo-cover-free, block-wise universal and the normal PRP assumption on block ciphers. Finally, we apply our main theorem to 3kf9 and PMAC_Plus, and successfully reduce their key sizes to the minimum possible. Thus, we solve a long-standing open problem in designing beyond-birthday-bound MAC with a single key.


Keywords: Beyond Birthday, 3kf9, PMAC_Plus, MAC, Sum of PRP, Coverfree, Rank, Strucutre Graph

## 1 Introduction

MAC. Message Authentication Code (MAC) is one of the important primitives in symmetric key cryptography to preserve the integrity of the message being transmitted. A MAC algorithm produces a fixed-length message digest, called a tag, from a variable-length message. For a secure MAC, it would be hard to forge a tag for a completely new message for which tag has not been observed. A stronger requirement of a MAC is pseudo-random function (PRF)
which informally says that the distribution of the tags be indistinguishable from uniform random distribution for any "efficient" adversary. The commonly used prf-advantage of a keyed construction $F$ against an adversary $\mathcal{A}$ is defined as follows:

$$
\operatorname{Adv}_{\mathbf{F}}^{\mathbf{p r f}}(\mathcal{A})=\operatorname{Pr}\left[\mathcal{A}^{F}=1\right]-\operatorname{Pr}\left[\mathcal{A}^{\Gamma}=1\right]
$$

where $\Gamma$ denotes the random function over the same domain and range as $F$. In practical applications, in addition to security, the issue of efficiency of MAC computation and the key-size are also very important.

Designing MAC/PRF. There are mainly three different approaches for designing a MAC: (a) universal hash function based, (b) compression function based, and (c) block cipher based. The drawback of universal hash based MAC design is that the performance of the MAC depends on the platform; some universal hash based MAC performs well in software, whereas the performance of others is noticeable only in hardware. In case of a compression function based MAC, the security of the MAC is established in terms of the prf-security of the underlying compression function. But designing a provably secure compression function got less attention than designing collision and (2nd) preimage compression function. On the other hand, analyzing block cipher becomes more popular.

Block Cipher Based MAC. Constructions that are based on block ciphers overcome the above difficulties. Performance of block cipher based MAC construction is balanced in both software and hardware. Examples of popular block cipher based MACs are CBC-MAC [3], OMAC [8], PMAC [5], TMAC [10] etc. However, for each of them, the so far best prf-security advantage is $O\left(L q^{2} / 2^{n}\right)$ where $q$ is the maximum number of queries, $L$ is maximum allowed message size and $n$ is the block-length of the underlying block cipher. For example, if PMAC is being implemented by PRINCE [5] (a 64-bit lightweight block cipher) in some small device and if we allow to encrypt up to $1 \mathrm{MB}\left(=2^{20}\right)$ messages then after 1 Million $\left(\approx 2^{20}\right)$ encryption, one may be able to distinguish it from random function with about $1 / 16$ probability. Thus, when $n$ is small (e.g., $<=64$ bits), as in lightweight cryptographic applications like RFIDs and smart cards, the birthday-bound security is no longer practical and so we need to seek for the construction achieving beyond birthday security.

Beyond-Birthday Secure MACs and Challenges. Among the block cipher based MACs that are beyond-birthday secure, two efficient (rate-1) constructions are PMAC_Plus and $3 \mathrm{kf9}$.

1. In CRYPTO 2011, Yasuda proposed PMAC_Plus, a simple three key variant of parallelizable and efficient PMAC. The author mentioned that - "This raises a challenge to come up with a 1-key rate-1 MAC construction which is secure beyond the birthday bound."
2. In ASIACRYPT 2012, Zhang et al proposed 3kf9 that improves the $f 9$ MAC mode adopted in the 3rd Generation Partnership Project (3GPP). 3kf9 also
requires three independent keys to lift its security beyond birthday-bound. Zhang et al. mentioned in their conclusion that, "However, its key size seems to be too large in some particular environments, requiring further improvements therefore."

There is also another deterministic MAC mode provides security beyond the birthday bound. As Dodis and Steinberger have shown, $\mathrm{MD}[f, g]$ reaches $O(\epsilon q$ poly $(n))$ MAC security. However, this design requires even longer keys and more block cipher invocations.

### 1.1 Motivation for Key-size Reduction in Block Cipher Based MACs

While beyond birthday bound block cipher modes are especially useful for smallsize block ciphers, their large key sizes prevent themselves from practical usages. This is more serious when implementing it in hardware, where registers to store key materials are expensive or otherwise injecting keys from outside brings security risks and slows down its overall efficiency. Furthermore, three block cipher keys imply three block cipher key schedulings, and this means, for most block ciphers (e.g. AES), three more block cipher invocation time and energy consumption.

A trivial way to reduce the key size, as commonly adopted in many practical protocols [1], is using a subkey generation algorithm $f$. Given a master key and some intermediate values, $f$ can derive several subkeys for each invocation on block ciphers. Despite of inconvenience, implementing and running $f$ requires extra memory and computation load, and its outputs pseudo-randomness is also a potential security risk, because to secure PMAC_Plus and 3 kf 9 we need three independently random keys.

A more technical method is to use tweakable block ciphers [13], which are expected to be independently random permutations with a single secret key and distinct-and-public tweaks. However, there are still some problems. If we adopt dedicated tweakable block ciphers, (e.g. [9]) in PMAC_Plus and 3kf9, we benefit from optimized efficiency but can hardly get provable security on normal block ciphers (PRP assumption); if we adopt birthday-bound tweakable block ciphers, e.g. [23, 7,17 ], we in fact lose the beyond-birthday bound in PMAC_Plus and 3kf9. Then we have to adopt the provably secure tweakable block ciphers with beyond-birthday-bound security, e.g. $[16,12,11,15]$. As far as we know, current solutions provide no good efficiency in our setting, because they need at least two normal block cipher invocations to build a tweakable block cipher, and their key sizes are not small either.

The Open Problem. Up to now, how to construct a beyond-birthday-bound MAC mode under a single key and reduce its security to the PRP assumption of its underlying block ciphers is still technically hard and remains as an open problem.

### 1.2 Our Contributions

With a view to solving the above problem, first we review the techniques used in3kf9 and PMAC_Plus. Despite their specific mechanisms to process message blocks, they both have doubled internal states sizes and then "encrypt" their last internal state by the well-known "Sum of PRPs". In proofs, cover-freeness of the final internal states is strictly necessary, and then by the previous results on "Sum of PRPs", the modes can reach a bound beyond the birthday paradox. With respect to the usages of key materials, the final "Sum of PRPs" needs two keys, and one more individual key is required by the message blocks processing phase. Then, if we just adopt a single key in these modes, we encounter two problems.

1. The first problem is "Sum of PRPs" may not work properly, because it will always output an all zero block once its two inputs collide.
2. The second problem is, the $q L$ block cipher inputs within internal structures may collide with the last two inputs ( $2 q$ in total) to "Sum of PRPs", and this seems to happen with a birthday bound probability. Once occurred, even though a specific attack is hard to present, a proof is however easily ruined because we can get no new randomness for the final output.

Obviously, designing a single-key such mode requires more techniques and its corresponding formal proof would be even harder and complex.

Contribution 1. To solve the first problem, we revisit the proofs for "Sum of PRPs", and propose a generalized but even simpler proof. Our basic observation is that the original provable security results hold even when the input domain is restricted. That is, over restricted domain and range, the sum of two same PRPs remains a PRF. Then we examine it by deriving "1-interpolation probability of sum of WOR (WithOut Replacement) samples", and generalize it to the $q$-interpolation case. As applications, we apply them to three/two-key sum constructions, and get successful proofs.

Contribution 2. To solve the second problem, we first define several notions, e.g. extended-cover-free, pseudo-cover-free, and block-wise universal, which are in fact abstracted from our analysis on one-key constructions. Taking advantages of this, we propose and prove our main theorem that can upper bound any one-key such construction by these items and an additional value.

Contributions 3 and 4. Finally, we turn to reduce the key size for $3 \mathrm{kf9}$ and PMAC_Plus. Taking advantage of our main theorem, in both proofs we just need to upper bound three items, i.e., extended-cover-free, pseudo-cover-free, and block-wise universal, for the corresponding CBC-like and PMAC-like structures. Though the proofs are more involved, interestingly, our obtained bounds for the one-key versions are only slightly larger than those for $3 \mathrm{kf9}$ and PMAC_Plus, and they essentially have the same form $O\left(q L_{\max } / 2^{n}+q^{3} L_{\max }^{3} / 2^{2 n}\right)$.

## 2 Preliminaries

Notation. We denote $X \stackrel{\$}{\stackrel{ }{\leftarrow}} S$ to mean that $X$ is chosen uniformly from $S$ and independently to all other random variables defined so far. We write $X \perp Y$ for independent random variables $X$ and $Y$. Let $[a . . b]:=\{a, a+1, \ldots, b\},[a]=$ [1..a]. By a $q$-set or $q$-tuple, we mean a set or a tuple of size $q$. Given a $q$-tuple $x=\left(x_{i}: i \in I\right)$, where $I$ is the index set, we abuse the notation $x$ also to mean the set $\left\{x_{i}: i \in I\right\}$. When all elements $x_{i}$ 's are distinct we simply write $x \in \operatorname{dist}_{q}$ or $x \in$ dist and we call $x$ element wise distinct. For a subset $J \subseteq I$, the sub-tuple $x_{J}:=\left(x_{j}\right)_{j \in J}$. In this paper, we fix a positive integer $n$ and all block ciphers considered in this paper have the block size $n$ and $L$ denotes the number of message blocks and $\ell$ denotes the length of the message it bits. Let $\mathbb{P}$ denote the set of all permutations over $\{0,1\}^{n}$. For any function $f$, and two tuples $x, y$ over same set of indices $I$, we write

$$
x \stackrel{f}{\longmapsto} y \text { to mean that } f\left(x_{i}\right)=y_{i}, \forall i \in I \text { and }
$$

Let $\mathbb{P}_{x \rightarrow y}:=\{\pi \in \mathbb{P}: x \stackrel{\pi}{\longmapsto} y\}$. For two tuples $x$ and $y$ over a same index set, we write $x \longrightarrow y$ (or $x \longleftrightarrow y$ ) if there exists a function (or permutation) respectively $\pi$ such that $x \stackrel{\pi}{\longmapsto} y$. In this case, we call $(x, y)$ function-compatible or (permutation-compatible) respectively.

### 2.1 Oracle Algorithm and Its Transcript

An oracle algorithm $\mathcal{A}$ (e.g., distinguisher or some block cipher based constructions in which block ciphers are viewed as oracles) interacting with one or more oracles $\mathcal{O}$ makes queries depending on the previous query responses. We denote the oracle interaction by $\mathcal{A}^{\mathcal{O}}(m)$ or $\mathcal{A}(m) \rightarrow \mathcal{O}$. During the interaction $\mathcal{A}^{\mathcal{O}}(m)$, let $X_{1}:=\left(X_{1,1}, \ldots, X_{1, r}\right)$ be the tuple of all queries to $\mathcal{O}$ and $Y_{1}:=\left(Y_{1,1}, \ldots, Y_{1, r}\right)$ be the tuple of corresponding responses. The transcript $\left(X_{1}, Y_{1}\right)$ is denoted as $\tau(\mathcal{A}(m) \rightarrow \mathcal{O})$. In case of a deterministic algorithm $\mathcal{A}$, $X_{1, i}$ is some function of $Y_{1,1}, \ldots, Y_{1, i-1}$ and $m$. Finally, it returns some output $c$ which must be a function of $Y$ and $m .{ }^{3}$ Let $\mathcal{A}$ be a deterministic oracle algorithm and a $q$-tuple $m=\left(m_{1}, \ldots, m_{q}\right)$. For any function $f$, we write the $q$-transcript of all query-responses $\left(X:=\left(X_{1}, \ldots, X_{q}\right), Y:=\left(Y_{1}, \ldots, Y_{q}\right)\right)$ as $\tau\left(\mathcal{A}(m) \rightarrow_{q} f\right)$ or simply as $\tau(\mathcal{A}(m) \rightarrow f)$ (whenever $q$ is understood from the context) by abusing notation where $\left(X_{i}, Y_{i}\right)=\tau\left(\mathcal{A}\left(m_{i}\right) \rightarrow f\right)$.

Definition 1. A pair of tuples $(x, y)$ is called $\mathcal{A}(m)$-realizable for a $q$-tuple $m$, if there exists a function $f$ such that $\tau\left(\mathcal{A}(m) \rightarrow_{q} f\right)=(x, y)$.

The following simple observation is very useful which abstracts a useful feature of query-responses for an interaction of a deterministic algorithm with a random function.

[^0]Lemma 1. Let $\mathcal{A}$ be a deterministic oracle algorithm. For any $\mathcal{A}(m)$-realizable pair $(x, y)$, we have $x \stackrel{f}{\longmapsto} y$ if and only if $\tau(\mathcal{A}(m) \rightarrow f)=(x, y)$. Thus, for any event $E$ and for any random function F ,

$$
\operatorname{Pr}_{\mathrm{F}}[E \mid \tau(\mathcal{A}(m) \rightarrow \mathrm{F})=(x, y)]=\operatorname{Pr}_{\mathrm{F}}[E \mid x \stackrel{\mathrm{~F}}{\longmapsto} y] .
$$

Proof. Clearly, if $\tau(\mathcal{A}(m) \longrightarrow f)=(x, y)$ then $x \stackrel{f}{\longmapsto} y$. Conversely, let $(X, Y)=$ $\tau(\mathcal{A}(m) \rightarrow f)$ then we can prove $X_{i}=x_{i}, Y_{i}=y_{i}$ by induction on the query index $i$. When $i=1, X_{1}=x_{1}$ since otherwise $(x, y)$ can not be realizable and so $Y_{1}=y_{1}$. Now suppose $X_{j}=x_{j}, Y_{j}=y_{j}$ for all $j<i$. As $\mathcal{A}$ is a deterministic algorithm $X_{i}=x_{i}$ (otherwise ( $x, y$ ) can not be realizable). So $Y_{i}=y_{i}$.

Random Functions. A random function is a function which is chosen from the set of all functions following some distribution. In particular, uniform random function, denoted $\Gamma_{n}$, (or uniform random permutation $\Pi_{n}$ ) is chosen uniformly from the set of all functions (or permutations respectively) from a specified finite domain to $\{0,1\}^{n}$.

Interpolation Probability. For any tuples $x, y$ with same index set, and a random function F we call $\operatorname{Pr}[x \stackrel{\mathrm{~F}}{\mapsto} y]$ interpolation probability. Let $x$ and $y$ be a tuple of elements from the domain and range of $\Gamma_{n}\left(\right.$ or $\left.\Pi_{n}\right)$ over a same set of indices. Moreover, let $s$ be the number of distinct elements in $x$. It is easy to see that the interpolation probability $\operatorname{Pr}\left[x \stackrel{\Gamma_{n}}{\mapsto} y\right]$ is positive and equals to $2^{-n s}$ if and only if $(x, y)$ function compatible. Similarly, $\operatorname{Pr}[x \stackrel{\Pi}{\mapsto} y]$ is positive and equals to $1 / P_{s}^{2^{n}}$ if and only if $(x, y)$ is permutation-compatible where $P_{s}^{N}:=N(N-1) \cdots(N-s+1)$. This observation can be extended to the conditional probability for the uniform random permutation. Let $((x, a),(y, b))$ be a permutation-compatible pair such that $a \cap x=\phi$ and $a \in \operatorname{dist}_{s}$ then

$$
\operatorname{Pr}\left[a \stackrel{\Pi_{r}}{\mapsto} b \mid x \stackrel{\Pi_{r}}{\mapsto} y\right] \geq 2^{-n s} .
$$

### 2.2 Security Definitions

Pseudorandom Function and Permutation. We define distinguishing advantage of an oracle algorithm $\mathcal{A}$ for distinguishing two random functions F from G as

$$
\begin{equation*}
\operatorname{Adv}_{\mathcal{A}}(\mathrm{F} ; \mathrm{G}):=\operatorname{Pr}\left[\mathcal{A}^{\mathrm{F}}=1\right]-\operatorname{Pr}\left[\mathcal{A}^{\mathrm{G}}=1\right] . \tag{1}
\end{equation*}
$$

We define prf-advantage and prp-advantage of $\mathcal{A}$ for an $n$-bit construction F respectively by

$$
\operatorname{Adv}_{\mathrm{F}}^{\mathrm{prf}}(\mathcal{A}):=\operatorname{Adv}_{\mathcal{A}}\left(\mathrm{F} ; \Gamma_{n}\right), \quad \operatorname{Adv}_{\mathrm{F}}^{\mathrm{prp}}(\mathcal{A})=\mathbf{A d v}_{\mathcal{A}}\left(\mathrm{F} ; \Pi_{n}\right)
$$

. By a $(q, \ell, t)$-distinguisher $\mathcal{A}$ we mean, $\mathcal{A}$ makes at most $q$ queries (querycomplexity) with at most $\ell$-bits in each query (data-complexity) and runs in time at most $t$ (time-complexity). One may include some other complexities,
e.g., memory complexity. We write $\operatorname{Adv}_{\mathrm{F}}^{\mathrm{xxx}}(q, \ell, t)=\max _{\mathcal{A}} \operatorname{Adv}_{\mathrm{F}}^{\mathrm{xxx}}(\mathcal{A})$ where maximum is taken over all ( $q, \ell, t$ )-distinguishers $\mathcal{A}$ and xxx denotes either prf or prp. A non-adaptive adversary fixes all its queries before it sees the responses.

Universal and Cover-Free. Now we define some other information-theoretic security advantages (in which there is no presence of an adversary). Let F be an $n$-bit random function then

$$
\operatorname{Adv}_{\mathrm{F}}^{\mathrm{univ}}(L \ell)=\max _{m_{1} \neq m_{2} \in\{0,1\} \leq \ell} \operatorname{Pr}\left[\mathrm{F}\left(m_{1}\right)=\mathrm{F}\left(m_{2}\right)\right]
$$

Let F be a random function which outputs two blocks, denoted $(\Sigma, \Theta) \in\left(\{0,1\}^{n}\right)^{2}$. For a $q$-tuple of distinct messages $m=\left(m_{1}, \ldots, m_{q}\right)$, we write $\mathrm{F}\left(m_{i}\right)=\left(\Sigma_{i}, \Theta_{i}\right)$. For a $q$-tuple of pairs $\left(\sigma_{i}, \theta_{i}\right)_{i}$, we say that

1. $\sigma_{i}$ (or $\theta_{i}$ ) is fresh if it is not same as $\sigma_{j}$ (or $\theta_{j}$ respectively) for some $j \neq i$.
2. We say that a tuple $\left(\sigma_{i}, \theta_{i}\right)_{i}$ is cover-free if for all $i$, either $\sigma_{i}$ or $\theta_{i}$ is fresh.

Definition 2. We define ( $q, L$ )-cover-free advantage as

$$
\mathbf{A d v}_{\mathrm{F}}^{\mathrm{cf}}(q, L)=\max _{m \in \text { dist }_{q}} \operatorname{Pr}\left[\left(\Sigma_{i}, \Theta_{i}\right)_{i} \text { is not cover-free }\right] .
$$

Clearly, $\boldsymbol{A d v}_{\mathrm{F}}^{\text {cf }}(q, L) \leq q^{3} \mathbf{A d v}_{\mathrm{F}}^{\text {cf }}(3, L)$. So it would be sufficient to concentrate on a triple of messages while bounding cover-free advantages. We say that a construction $F$ is $(q, L, t, \epsilon)$ - $\operatorname{xxx}$ if $\mathbf{A d v}_{\mathrm{F}}^{\mathrm{xxx}}(q, L, t) \leq \epsilon$ where $\operatorname{xxx}$ denotes either univ or cf.

### 2.3 Coefficient H-Technique

In this section we discuss briefly Coefficient-H Tehcnique [20] which is also known as Decorrelation Theorem due to Vaudenay [25].
Definition 3 (statistical distance). Let $X$ and $Y$ two random variables over a set $S$. We define the statistical distance between $X$ and $Y$ as

$$
\Delta(X ; Y)=\max _{T \subseteq S} \operatorname{Pr}[X \in T]-\operatorname{Pr}[Y \in T] .
$$

We write $X \succ_{\epsilon} Y$ if $\operatorname{Pr}[X=s] \geq(1-\epsilon) \times \operatorname{Pr}[Y=s], \forall s$ and we say that $X \succ_{\epsilon} Y$ over $E$, if this holds only for all $s \in E$. We state a tool which would be used to bound the statistical distance between two random variables. The coefficient H-technique is the generalized version of this result for bounding distinguishing advantage of two random systems or probabilistic oracles.

Lemma 2 (coefficient $\mathbf{H}$-technique for random variables). Let $X, Y$ be two random variables over $S$ such that $X \succ_{\epsilon} Y$ over $\mathcal{V}_{\text {good }} \subseteq S$ then,

$$
\Delta(X ; Y) \leq \epsilon+\operatorname{Pr}\left[Y \notin \mathcal{V}_{\text {good }}\right]
$$

Proof. Let $T \subseteq S$. Then, $X \succ_{\epsilon} Y$ over $\mathcal{V}_{\text {good }}$ implies that

$$
\operatorname{Pr}\left[Y \in \mathcal{V}_{\text {good }} \cap T\right]-\operatorname{Pr}\left[X \in \mathcal{V}_{\text {good }} \cap T\right] \leq \epsilon \times \operatorname{Pr}\left[Y \in \mathcal{V}_{\text {good }} \cap T\right] \leq \epsilon
$$

So,

$$
\begin{aligned}
\operatorname{Pr}[Y \in T]-\operatorname{Pr}[X \in T] & \leq \epsilon+\left(\operatorname{Pr}\left[Y \in T \backslash \mathcal{V}_{\text {good }}\right]-\operatorname{Pr}\left[X \in T \backslash \mathcal{V}_{\text {good }}\right]\right) \\
& \leq \epsilon+\operatorname{Pr}\left[Y \notin \mathcal{V}_{\text {good }}\right]
\end{aligned}
$$

Hence the result follows.
Theorem 1 (coefficient H -technique for random functions). Let F and G be two random functions. Let $\mathcal{V}_{\text {good }} \subseteq \mathcal{X}^{q} \times \mathcal{Y}^{q}$. If (i) for all $q$-distinct messages $m=\left(m_{1}, \ldots, m_{q}\right),\left(\mathrm{F}\left(m_{i}\right)\right)_{i} \succ_{\epsilon_{1}}\left(\mathrm{G}\left(m_{i}\right)\right)_{i}$ over $\mathcal{V}_{\text {good }}$ (ii) $\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right) \notin \mathcal{V}_{\text {good }}\right] \leq \epsilon_{2}$ then $\mathbf{A d v}_{\mathcal{A}}(\mathrm{F} ; \mathrm{G}) \leq \epsilon_{1}+\epsilon_{2}$.

Proof. Condition (i) says that, for all $v \in V_{\text {good }}, \operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right)=v\right]-\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=\right.$ $v] \leq \epsilon_{1} \cdot \operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=v\right]$. Now,

$$
\begin{aligned}
\operatorname{Adv}_{\mathcal{A}}(\mathrm{F} ; \mathrm{G})= & \operatorname{Pr}\left[\mathcal{A}^{\mathrm{G}}=1\right]-\operatorname{Pr}\left[\mathcal{A}^{\mathrm{F}}=1\right] \\
= & \sum_{v \in V}\left(\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right)=v\right]-\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=v\right]\right) \\
= & \sum_{v \in V \cap V_{\text {good }}}\left(\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right)=v\right]-\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=v\right]\right) \\
& +\sum_{v \notin V_{\text {good }}}\left(\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right)=v\right]-\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=v\right]\right) \\
\leq & \left(\sum_{v \in V \cap V_{\text {good }}} \epsilon_{1} \cdot \operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{F}}\right)=v\right]\right)+\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\mathrm{G}}\right) \notin V_{\text {good }}\right] \\
\leq & \epsilon_{1}+\epsilon_{2}
\end{aligned}
$$

## 3 New Proposals for Beyond-Birthday Secure One Key MAC

We introduce here the construction of two separate MACs. One is 1 kf 9 MAC and another is 1 k -PMAC+ both of the constructions require a single key $K$.

### 3.1 1kf9-MAC

In this section we present the algorithm for 1 kf 9 MAC followed by its schematic diagram. For any message $M \in\{0,1\}^{*}, 1 \mathrm{kf} 9$ Algorithm first prepends a all zero block to the message and pads it to make the length multiple of the block length $n$. Then $M$ is iteratively processed through block cipher $E_{K}$ as shown in Fig. 3.1 and the final tag $T$ is obtained by XOR-ing $E_{K}(\Sigma)$ and $E_{K}(\Theta)$.

```
    Input: \(K \stackrel{\$}{\leftarrow} \mathcal{K}, M \leftarrow\{0,1\}^{*}\)
    Output: \(T \in\{0,1\}^{n}\)
    \(M \leftarrow 0^{n}| | M| | 10^{n-1-l e n(M) \bmod n}\);
    \(M_{1}, M_{2}, \ldots, M_{l} \leftarrow \operatorname{Partition}(M) ;\)
    \(Y_{0} \leftarrow E_{K}\left(0^{n}\right) ;\)
    \(Z \leftarrow Y_{0} ;\)
    for \(j=1\) to \(l\) do
        \(X_{j}=Y_{j-1} \oplus M_{j} ;\)
        \(Y_{j}=E_{K}\left(X_{j}\right)\);
        \(Z=\left(Z \oplus Y_{j}\right) ;\)
    end
    \(\Theta^{\prime}=2 Z ;\)
\(10 \Sigma^{\prime}=2 Y_{l}\);
\(11 \Sigma=\) fix \(0\left(\Sigma^{\prime}\right)\);
\(12 \Theta=\) fix1 \(\left(\Theta^{\prime}\right)\);
\(13 T \leftarrow E_{K}(\Sigma) \oplus E_{K}(\Theta)\);
14 Return \(T\);
```

Algorithm 1: Algorithm of 1kf9-MAC


Fig. 3.1. Construction of 1kf9-MAC

### 3.2 1k-PMAC+

In this section we present the algorithm for 1 k -PMAC+ followed by its schematic diagram. $\quad \Delta_{i}$ is the encryption of field element $\mathrm{Cst}_{i}$ for $i=1,2$. After suitable

```
    Input: \(K \stackrel{\&}{\leftarrow} \mathcal{K}, M \leftarrow\{0,1\}^{*}\)
    Output: \(T \in\{0,1\}^{n}\)
    \(\Delta_{i} \leftarrow E_{K}\left(\mathrm{Cst}_{i}\right)\) for \(i=1,2\);
    \(M \leftarrow M \| 10^{n-1-l e n(M) \bmod n}\);
    \(M_{1}, M_{2}, \ldots, M_{l} \leftarrow \operatorname{Partition}(M) ;\)
    for \(j=1\) to \(l\) do
        \(X_{j}=M_{j} \oplus 2^{j-1} \cdot \Delta_{1} \oplus 2^{2(j-1)} \cdot \Delta_{2} ;\)
        \(Y_{j}=E_{K}\left(X_{j}\right) ;\)
    end
    \(\Sigma^{\prime}=Y_{1} \oplus Y_{2} \oplus \ldots \oplus Y_{l} ;\)
    \(\Theta^{\prime}=2^{l} \cdot Y_{1} \oplus 2^{l-1} \cdot Y_{2} \oplus \ldots \oplus 2 \cdot Y_{l} ;\)
    \(\Sigma=\operatorname{fix} 0\left(\Sigma^{\prime}\right)\);
    \(\Theta=\mathrm{fix} 1\left(\Theta^{\prime}\right)\);
\(10 T \leftarrow E_{K}(\Sigma) \oplus E_{K}(\Theta)\);
11 Return \(T\);
```

Algorithm 2: Algorithm of 1-Key PMAC+
padding of the message $M$, each block is processed in parallel fashion as shown in Fig. 3.2. $\Sigma^{\prime}$ is obtained by sum of the all the intermediate outputs and $\Theta^{\prime}$ is obtained by a linear combination of the intermediate outputs. $\Sigma$ is obtained by fix 0 on $\Sigma^{\prime}$ and $\Theta$ is obtained by fix 1 on $\Theta^{\prime}$. Then the xor of $E_{K}(\Sigma)$ and $E_{K}(\Theta)$ is returned as the tag $T$ of message $M$.


Fig. 3.2. Construction of 1Key PMAC+

### 3.3 Comparison Chart of Our Construction with 3kf9 and PMAC+

| Construction | Reference | No.of Keys Required | Security Bound |
| :---: | :---: | :---: | :---: |
| Sum of CBC | $[26]$ | 4-Keys | $O\left(l^{4} q^{3} / 2^{2 n}\right) / O\left(l^{3} q^{3} / 2^{2 n}\right)$ |
| PMAC+ | $[27]$ | 3-keys | $O\left(l^{3} q^{3} / 2^{2 n}+l q / 2^{n}\right)$ |
| 3kf9 | $[28]$ | 3-keys | $O\left(l^{3} q^{3} / 2^{2 n}+l q / 2^{n}\right)$ |
| 1kf9 | This Paper | 1-key | $O\left(q l^{2} / 2^{n}+q^{3} l^{4} / 2^{2 n}+q^{4} l^{4} / 2^{3 n}+q^{4} l^{6} / 2^{4 n}\right)$ |
| 1k-PMAC+ | This Paper | 1-key | $O\left(q l^{2} / 2^{n}+q^{3} l^{4} / 2^{2 n}+q^{4} l^{4} / 2^{3 n}+q^{4} l^{6} / 2^{4 n}\right)$ |

### 3.4 Design Rationale

In 1 kf 9 , we prepend a 0 block message to ensure that $\Sigma, \Theta \neq 0$. Moreover, in order to ensure that $\Sigma \neq \Theta$, we fix the last bit of $\Sigma^{\prime}$ to 0 and that of $\Theta^{\prime}$ to 1.To ensure the deired rank as described in Section 9, we multiply the intermediate output with 2 and add them.

In 1 k -PMAC+ we use a double mask that ensures the rank of bad equations described in Section 9 is at least 2 and we use fix0 and fix1 to ensure $\Sigma \neq \Theta$.

## 4 Some Results on Sampling Distributions

### 4.1 With (out) replacement sampling

Let $\left(Y_{1}, \ldots, Y_{r}\right) \stackrel{\text { wor }}{\leftarrow} S$ be a set of $r$ samples drawn without replacement from a set $S$. In other words, the conditional distribution

$$
Y_{i} \mid\left(Y_{1}, \ldots, Y_{i-1}\right) \stackrel{\$}{\leftarrow} S \backslash\left\{Y_{1}, \ldots, Y_{i-1}\right\} .
$$

Similarly, for the with replacement sampling, we write $U:=\left(U_{1}, \ldots, U_{r}\right) \stackrel{\text { wr }}{\leftarrow} S$ which is same as drawing $U_{i}$ 's uniformly and independently from the set $S$. Let us consider the following question.

## How close the sum of two WOR sampling to WR ?

More precisely, let $U:=\left(U_{1}, \ldots, U_{q}\right) \stackrel{\text { wr }}{\leftarrow}\{0,1\}^{n}$. We would like to obtain an upper bound of the statistical distance

$$
\Delta\left(\left(Z_{1}, \ldots, Z_{q}\right) ;\left(U_{1}, \ldots, U_{q}\right)\right)
$$

where $Z_{i}=Y_{1, i} \oplus Y_{2, i}, 1 \leq i \leq q$, and the joint distributions of $Y$ 's are any one of the followings cases.

Case-1 (sum of two independent WOR over $\left.\{0,1\}^{n}\right): Y_{1}=\left(Y_{1,1}, \ldots, Y_{1, q}\right) \stackrel{\text { wor }}{\leftarrow}\{0,1\}^{n}$ and $Y_{2}=\left(Y_{2,1}, \ldots, Y_{2, q}\right) \stackrel{\text { wor }}{\leftarrow}\{0,1\}^{n}$ and $Y_{1} \perp Y_{2}$.

Case-2 (sum of two dependent WOR over $\left.\{0,1\}^{n}\right):\left(Y_{1,1}, Y_{2,1}, \ldots, Y_{1, q}, Y_{2, q}\right) \stackrel{\text { wor }}{\leftarrow}$ $\{0,1\}^{n}$.
Case-3 (sum of two dependent WOR over $S$ ): $\left(Y_{1,1}, Y_{2,1}, \ldots, Y_{1, q}, Y_{2, q}\right) \stackrel{\text { wor }}{\leftarrow} S \subseteq$ $\{0,1\}^{n}$ for a set $S$ with size $2^{n}-\sigma, \sigma \geq 0$.

For the first two cases, Bellare et.al [2] had shown that $\Delta(Z ; U) \leq$ $\frac{q}{2^{n}}+\mathcal{O}\left(n \times\left(\frac{q}{2^{n}}\right)^{1.5}\right)$. Their analysis uses some advanced results of probability theory (e.g., Azuma's inequality and Chernoff theorem). For the first case, later Lucks [14] provided an elementary proof with the upper bound $\mathcal{O}\left(q^{3} / 2^{2 n}\right)$ and Patarin [22] provides a much involved complex proof with the upper bound $\mathcal{O}\left(q / 2^{n}\right)$.

In this paper, we consider the third case which also generalizes case 2 when $S=\{0,1\}^{n}$. Our analysis is similar to that of Lucks [14] but much more simplified and can be similarly applicable to the first case. In the next section, we see the application of this result for analyzing one-key constructions of a specific form. We now state the key lemma which would be used to bound the statistical distance between sum of WOR sampling and WR sampling.

Lemma 3 (1-interpolation probability of sum of WOR samples). Let $S^{\prime} \subseteq\{0,1\}^{n}$ be a subset of size $\left(2^{n}-s^{\prime}\right)$ and $U_{n} \stackrel{\$}{\leftarrow}\{0,1\}^{n}$. Let $(V, W) \stackrel{\text { wor }}{\leftarrow} S^{\prime}$ be a WOR sample of size 2 drawn from $S^{\prime}$. Then, $V \oplus W \succ_{\epsilon} U_{n}$ over $\mathbb{F}_{2^{n}}^{*}:=\mathbb{F}_{2^{n}} \backslash\left\{0^{n}\right\}$ where $\epsilon:=\frac{s^{\prime 2}}{\left(2^{2}-s^{\prime}\right)^{2}}$.

Proof. Let $t \in \mathbb{F}_{2^{n}}^{*}$. For $i=1,2$, let $A_{i}=\left\{\left(a_{1}, a_{2}\right): a_{1} \oplus a_{2}=t, a_{i} \notin S^{\prime}\right\}$. Clearly, $\left|A_{i}\right| \leq s^{\prime}$. Note that $\left\{(x, y) \in S^{\prime} \times S^{\prime}: x \oplus y=t\right\}=\{(x, t \oplus x): x \in$ $\left.\{0,1\}^{n}\right\} \backslash\left(A_{1} \cup A_{2}\right)$. So,

$$
\begin{aligned}
\operatorname{Pr}[V \oplus W=t] & =\frac{2^{n}-\left|A_{1} \cup A_{2}\right|}{\left(2^{n}-s^{\prime}\right)\left(2^{n}-s^{\prime}-1\right)} \\
& \geq \frac{2^{n}-2 s^{\prime}}{\left(2^{n}-s^{\prime}\right)^{2}}=2^{-n}\left(1-\frac{s^{\prime 2}}{\left(2^{n}-s^{\prime}\right)^{2}}\right) .
\end{aligned}
$$

Observation 1. The above result is also valid if $(V, W) \stackrel{\$}{\leftarrow} S^{\prime} \times T^{\prime}$ such that $\left|S^{\prime}\right|=\left|T^{\prime}\right|=2^{n}-s^{\prime}$. Then, exactly same argument and hence result holds (i.e., $V \oplus W \succ_{\epsilon} U_{n}$ over $\mathbb{F}_{2^{n}}^{*}$. When $s^{\prime} \leq 2^{n-1}, \epsilon \leq 4 s^{\prime 2} / 2^{2 n}$.

Theorem 2 ( $q$-interpolation probability of sum of dependent WOR samples over $S$ ). Let $S \subseteq\{0,1\}^{n}$ of size $2^{n}-s$, $\left(Y_{1,1}, Y_{2,1}, \ldots, Y_{1, q}, Y_{2, q}\right) \stackrel{\text { wor }}{\leftarrow} S$ and let $Z=\left(Z_{1}:=\left(Y_{1,1} \oplus Y_{2,1}\right), \ldots, Z_{q}:=\left(Y_{1, q} \oplus Y_{2, q}\right)\right)$. Then,

$$
Z \succ_{\epsilon} U \text { over } \mathbb{F}_{2^{n}}^{*} \text { where } \epsilon:=\frac{q s^{2}+2 s q^{2}+4 q^{3} / 3}{\left(2^{n}-s-2 q\right)^{2}}
$$

Proof. Let $S^{c}=\left\{a_{0}, a_{-1}, \ldots, a_{-s+1}\right\}$. Let us fix $i \geq 1, t=\left(t_{1}, \ldots, t_{q}\right) \in\left(\mathbb{F}_{2^{n}}^{*}\right)^{q}$ and $a_{1}, a_{2}, \ldots, a_{2 i-3}, a_{2 i-2}$ be distinct elements from $S$ such that $a_{2 j-1} \oplus a_{2 j}=t_{j}$, $1 \leq j<i$. By using Lemma 3 with $S^{\prime}=\{0,1\}^{n} \backslash\left\{a_{j}:-s<j \leq 2 i-2\right\}$ and $s^{\prime}=s+2(i-1)$, we have
$\operatorname{Pr}\left[Z_{i}=t_{i} \mid Y_{1,1}=a_{1}, Y_{2,1}=a_{2}, \ldots, Y_{1, i-1}=a_{2 i-3}, Y_{2, i-1}=a_{2 i-2}\right] \geq \frac{1}{2^{n}}\left(1-\epsilon_{i}\right)$
where $\epsilon_{i}=\frac{(s+2(i-1))^{2}}{\left(2^{n}-s-2(i-1)\right)^{2}}$. Since this bound holds for any $a_{i}$ 's, we can conclude that $\operatorname{Pr}\left[Z_{i}=t_{i} \mid Z_{1}=t_{1}, \ldots, Z_{i-1}=t_{i-1}\right] \geq \frac{1}{2^{n}}\left(1-\epsilon_{i}\right)$. After applying chain rule for these conditional probabilities, we obtain that

$$
\begin{equation*}
\operatorname{Pr}[Z=t] \geq 2^{-n q}\left(1-\sum_{i} \epsilon_{i}\right) \geq 2^{-n q}\left(1-\frac{q s^{2}+2 s q^{2}+4 q^{3} / 3}{\left(2^{n}-s-2 q\right)^{2}}\right) \tag{2}
\end{equation*}
$$

Observation 2. Same argument also works when $Y_{1}:=\left(Y_{1,1}, Y_{1,2} \ldots, Y_{1, q}\right)$ wor $S, Y_{2}:=\left(Y_{2,1}, Y_{2,2}, \ldots, Y_{2, q}\right) \stackrel{\text { wor }}{\leftarrow} T$ are two $q$-samples and $Y_{1} \perp Y_{2}$ where $S$ and $T$ are two subsets of $\{0,1\}^{n}$ of size $2^{n}-s$. On the calculation of the conditional probability of $Z_{i}$, we set $S^{\prime}=\{0,1\}^{n} \backslash\left(S^{c} \cup\left\{a_{1}, a_{3}, \ldots, a_{2 i-3}\right\}\right)$ and $T^{\prime}=$ $\{0,1\}^{n} \backslash\left(T^{c} \cup\left\{a_{2}, a_{4}, \ldots, a_{2 i-2}\right\}\right)$ and so we set $s_{i}^{\prime}=s+(i-1)$. Then using our Observation 1, the Equation (2) holds with $\epsilon_{i}=s_{i}^{\prime 2} /\left(2^{n}-s_{i}^{\prime}\right)^{2}$. After simplifying $\sum_{i} \epsilon_{i}$, we can conclude that $Z \succ_{\epsilon} U$ where $\epsilon=\frac{q s^{2}+s q^{2}+q^{3} / 3}{\left(2^{n}-s-q\right)^{2}} \leq \frac{4 q s^{2}+4 s q^{2}+4 q^{3} / 3}{2^{2 n}}$ provided $s+q<2^{n-1}$.

Now we summarize our results in the view of all cases we initially aimed to answer. We denote $Z_{i}=X_{i} \oplus Y_{i}$ and $Z=\left(Z_{1}, \ldots, Z_{q}\right)$ and $U:=\left(U_{1}, \ldots, U_{q}\right) \stackrel{\text { wr }}{\leftarrow}$ $\{0,1\}^{n}$.

Corollary 1. Let $X \stackrel{\text { wor }}{\leftarrow} S$ and $Y \stackrel{\text { wor }}{\leftarrow} T$ be two independent $q$-samples such that $S, T \subseteq\{0,1\}^{n}$ of size $2^{n}-s$. If $s \leq 2^{n-1}-q$ then

$$
\Delta(Z ; U) \leq \frac{q}{2^{n}}+\frac{4 q s^{2}+4 s q^{2}+4 q^{3} / 3}{2^{2 n}}
$$

over $\mathbb{F}_{2^{n}}^{*}$.
In particular, for Case-1 we have $S=T=\{0,1\}^{n}$ (i.e., $s=0$ ) and so $\Delta(Z ; U) \leq \frac{q}{2^{n}}+\frac{4 q^{3} / 3}{2^{2 n}}$ over $\mathbb{F}_{2^{n}}^{*}$.

Corollary 2. Let $\left(X_{1}, Y_{1}, \ldots, X_{q}, Y_{q}\right) \stackrel{\text { wor }}{\leftarrow} S \subseteq\{0,1\}^{n}$ such that $\left|S^{c}\right|:=s \leq$ $2^{n-1}-2 q$. Then (a) in (Case-3),

$$
\Delta(Z ; U) \leq \frac{q}{2^{n}}+\frac{4 q s^{2}+8 s q^{2}+6 q^{3}}{2^{2 n}}
$$

over $\mathbb{F}_{2^{n}}^{*}$.
If in addition if $q \leq s$ then $\Delta(Z ; U) \leq \frac{q}{2^{n}}+\frac{18 s^{3}}{2^{2 n}}$. (b) For $s=0$ (i.e, in Case-2) we have $\Delta(Z ; U) \leq \frac{q}{2^{n}}+\frac{6 q^{3}}{2^{2 n}}$ over $\mathbb{F}_{2^{n}}^{*}$.

### 4.2 Applications to PRF Security of Sum of Uniform Random Permutation

Let $\Pi$ be a uniform random permutation on $\{0,1\}^{n}$. Then, for any distinct $x_{1}, \ldots, x_{q}$, it is easy to see that $\Pi^{(q)}(x):=\left(\Pi\left(x_{1}\right), \ldots, \Pi\left(x_{q}\right)\right) \stackrel{\text { wor }}{\leftarrow}\{0,1\}^{n}$. So when $\Pi_{1}$ and $\Pi_{2}$ are two independent uniform random permutations then, $\Pi_{1}^{(q)}(x) \stackrel{\text { wor }}{\leftarrow}\{0,1\}^{n}, \Pi_{2}^{(q)}(x) \stackrel{\text { wor }}{\leftarrow}\{0,1\}^{n}$ and $\Pi_{1}^{(q)}(x) \perp \Pi_{2}^{(q)}(x)$ where $x \in$ dist.
Case-a. The Case-1 actually talks about the pseudorandomness of sum of two independent random permutations. More precisely, let $\mathrm{SUM}_{1}^{\Pi_{1}, \Pi_{2}}(x)=\Pi_{1}(x) \oplus$ $\Pi_{2}(x)$ where $\Pi_{1}$ and $\Pi_{2}$ are two independent random permutations. Then, using Corollary 1, we have

$$
\operatorname{Adv}_{\operatorname{SuM}_{1}^{\Pi_{1}, \Pi_{2}}}^{\operatorname{prf}}(q) \leq \frac{q}{2^{n}}+\frac{4 q^{3} / 3}{2^{2 n}}
$$

Case-b. Case-2 talks about the pseudorandomness of $\left(\Pi\left(x_{1}\right) \oplus \Pi\left(x_{2}\right), \ldots\right.$, $\left.\Pi\left(x_{2 q-2}\right) \oplus \Pi\left(x_{2 q}\right)\right)$ where $x=\left(x_{1}, \ldots, x_{2 q}\right)$ is element wise distinct. We can define a function $\operatorname{SUM}_{2}^{\Pi}:\{0,1\}^{n-1} \rightarrow\{0,1\}^{n}$ mapping an $(n-1)$ bit string $y$ to $\Pi(0 \| y) \oplus \Pi(1 \| y)$. So using (b) of Corollary 2 we have,

$$
\mathbf{A d v}_{\mathrm{SUM}_{2}^{\Pi}}^{\mathrm{prf}}(q) \leq \frac{q}{2^{n}}+\frac{6 q^{3}}{2^{2 n}}
$$

The above construction has been analyzed in [2].
Case-c. Now we come to the Case-3 which deals with the pseudorandomness of $\left(\Pi^{*}\left(x_{1}\right) \oplus \Pi^{*}\left(x_{2}\right), \ldots, \Pi^{*}\left(x_{2 q-2}\right) \oplus \Pi^{*}\left(x_{2 q}\right)\right)$ where $\Pi^{*} \stackrel{\$}{\leftarrow} \mathbb{P}_{a \rightarrow b}$ for two element wise distinct $s$-tuples $a, b$, and $x \cap a=\phi$. Suppose we restrict the domain of $\operatorname{SUM}_{2}^{\Pi *}$ (as defined above) to $D:=\left\{y \in\{0,1\}^{n-1}: 0\|y, 1\| y \notin a\right\}$. Then, for all $q \leq s$, we have

$$
\operatorname{Adv}_{\mathrm{SUM}_{2}^{\Pi *}}^{\mathrm{prf}}(q) \leq \frac{q}{2^{n}}+\frac{18 s^{3}}{2^{2 n}}
$$

We also state a theorem involving interpolation probability which would be used later for prf security analysis of sum-based construction. The proof of the theorem is obvious from (a) of Corollary 2.

We define sum function over two blocks as follows: $\operatorname{sum}^{\pi}(x, y)=\pi(x) \oplus \pi(y)$ and $\operatorname{sum}^{\pi_{1}, \pi_{2}}(x, y)=\pi_{1}(x) \oplus \pi_{2}(y)$

Theorem 3. Let $(x, y)$ be a permutation compatible pair of $s$-tuples. Let $\sigma_{1}, \theta_{1}$, $\ldots, \sigma_{q}, \theta_{q}$ be $2 q$ distinct elements from the set $\{0,1\}^{n} \backslash x$. If $s+2 q \leq 2^{n-1}$ then, for any non-zero $t_{1}, \ldots, t_{q} \in\{0,1\}^{n}$,

$$
\frac{1}{\left(2^{n}-s\right)^{q}} \geq \operatorname{Pr}\left[\left(\sigma_{i}, \theta_{i}\right)_{i} \stackrel{\text { sum }}{\longrightarrow} t \mid x \stackrel{\Pi}{\longmapsto} y\right] \geq 2^{-n q}(1-\epsilon)
$$

where $\epsilon=\frac{4 q s^{2}+8 s q^{2}+6 q^{3}}{2^{2 n}}$.

Proof. Set $Y_{1, i}=\Pi_{x \rightarrow y}\left(\sigma_{i}\right), Y_{2, i}=\Pi_{x \rightarrow y}\left(\theta_{i}\right)$ then $\left(Y_{1}, Y_{2}\right) \stackrel{\text { wor }}{\leftarrow} S:=\{0,1\}^{n} \backslash y$. Hence we can apply Theorem 2 to conclude our theorem.

A simpler version of the above theorem when $s=0$ and we consider sum of two uniform random permutations, we have the following result. The proof is again straightforward from Observation 2.

Theorem 4. Let $(x, y)$ and $\left(x^{\prime}, y^{\prime}\right)$ be two permutation compatible pair of $s$ tuples. Let $\sigma_{1}, \ldots \sigma_{q}$ be $q$ distinct elements from the set $\{0,1\}^{n} \backslash x$ and $\theta_{1}, \ldots \theta_{q}$ be $q$ distinct elements from the set $\{0,1\}^{n} \backslash x^{\prime}$. If $s+q \leq 2^{n-1}$ then, for any non-zero $t_{1}, \ldots, t_{q} \in\{0,1\}^{n}, \frac{1}{\left(2^{n}-s\right)^{q}} \geq \operatorname{Pr}\left[\left(\sigma_{i}, \theta_{i}\right)_{i} \xrightarrow{\text { sum }^{\Pi_{1}, \Pi_{2}}} t \mid x \xrightarrow{\stackrel{\Pi_{1}}{\longmapsto}} y, x^{\prime} \xrightarrow{\Pi_{1}}\right.$ $\left.y^{\prime}\right] \geq 2^{-n q}\left(1-\frac{4 q s^{2}+4 s q^{2}+4 q^{3} / 3}{2^{2 n}}\right)$.

## 5 A Generic Hash-then-Sum Construction

An affine mode is a deterministic oracle algorithm whose query computations (functions) are affine functions and its oracle is some random function.
Block-Separated Double Block Construction. Let $\mathcal{C}^{\pi}:\{0,1\}^{*} \rightarrow R$ be a permutation-based deterministic construction. When $e$ is a blockcipher then for any key $K, e_{K}$ is an $n$-bit permutation. Thus, a blockcipher based construction $\mathcal{C}^{{ }^{e}}{ }_{K}$ can be viewed as a permutation-based construction $\mathcal{C}^{\pi}$. When $R=\{0,1\}^{2 n}$, it is called a double block construction and we write the two output blocks as $\mathcal{C}^{\pi}(m)=(\Sigma, \Theta)$. We say that $\mathcal{C}$ is block-separated if the range of possible values of $\Sigma$ and $\Theta$ are disjoint. More formally, for all $m_{1} \neq m_{2}$, and for all permutation $\pi$ if

$$
\mathcal{C}^{\pi}\left(m_{1}\right)=\left(\Sigma_{1}, \Theta_{1}\right), \mathcal{C}^{\pi}\left(m_{2}\right)=\left(\Sigma_{2}, \Theta_{2}\right) \Rightarrow \Sigma_{1} \neq \Theta_{2}
$$

For any double construction, with a minor modification, one can make it blockseparated. For example, let fix0 : $\{0,1\}^{n} \rightarrow\{0,1\}^{n}$ be a function mapping $x_{1} x_{2} \cdots x_{n}$ to $0 x_{2} \cdots x_{n}$. Similarly, we define fix1 which fixes the first bits to 1 . Now, the double block construction defined as $\mathcal{C}^{\prime}=\left(\Sigma^{\prime}, \Theta^{\prime}\right)$ is block-separated where $\Sigma^{\prime}=\mathrm{fix} 0(\Sigma)$ and $\Theta^{\prime}=\mathrm{fix} 1(\Theta)$.
A Composition Theorem: $\operatorname{PRF}(\mathrm{U}) \equiv$ PRF. It is well known [24] that composition of $\epsilon$ universal hash function $\mathcal{H}$ and a PRF $g$ is a PRF which has been proved using game-playing technique. For the sake of completeness, we formally prove the theorem using Patarin's Coefficient-H Techhnique.

Theorem 5. Let $F_{K_{1}, K_{2}}:=g_{K_{2}} \circ \mathcal{H}_{K_{1}}:\{0,1\}^{*} \rightarrow\{0,1\}^{n}$. Then,

$$
\mathbf{A d v}_{F}^{\mathrm{prf}}(q, \ell, t) \leq \mathbf{A d v}_{g}^{\mathrm{prf}}\left(q, \ell, t^{\prime}\right)+\binom{q}{2} \times \mathbf{A d v}_{\mathcal{H}}^{\mathrm{univ}}(\ell)
$$

where $t^{\prime}=t+\mathcal{O}\left(q T_{\ell}\right)$ and $T_{\ell}$ denotes the maximum time for computing $\mathcal{H}(m)$ for any $\ell$-bit message $m$.

Proof. For the sake of completeness, we quickly revise the proof of the statement by using coefficient H -technique. By using standard reduction argument, we can consider the composition function $\Gamma_{n} \circ \mathcal{H}_{K_{1}}$ at the cost of $\mathbf{A d v}_{g}^{\text {prf }}\left(q, \ell, t^{\prime}\right)$. Now, for any $q$-tuple $m=\left(m_{1}, \ldots, m_{q}\right)$ of distinct messages, we denote $\mathcal{H}_{K_{1}}\left(m_{i}\right)=X_{i}$. For all $t=\left(t_{1}, \ldots, t_{q}\right) \in\left(\{0,1\}^{n}\right)^{q}$, the interpolation probability

$$
\begin{aligned}
\operatorname{Pr}_{\Gamma_{n}, K_{1}}\left[m \stackrel{\Gamma_{n} \circ \mathcal{H}_{K_{1}}}{\longmapsto} t\right] & \geq \sum_{x \in \text { dist }_{q}} \operatorname{Pr}\left[x \stackrel{\Gamma_{n}}{\longrightarrow} t \mid X=x\right] \times \operatorname{Pr}[X=x] \\
& =2^{-n q} \times \operatorname{Pr}\left[X \in \operatorname{dist}_{q}\right] \\
& \geq 2^{-n q} \times\left(1-\sum_{1 \leq i<j \leq q} \operatorname{Pr}\left[X_{i}=X_{j}\right]\right) \\
& \geq 2^{-n q} \times\left(1-\binom{q}{2} \mathbf{A d v}_{\mathcal{H}}^{\text {univ }}(\ell)\right) .
\end{aligned}
$$

Beyond Birthday Security. To achieve the beyond birthday security, one can consider $\mathcal{H}_{K_{1}}:\{0,1\}^{*} \rightarrow\{0,1\}^{2 n}$ and $g_{K_{2}}:\{0,1\}^{2 n} \rightarrow\{0,1\}^{n}$. So if $\operatorname{Adv}_{\mathcal{H}}^{\text {univ }}(\ell)=O\left(2^{-2 n}\right)$ and $g$ has beyond birthday prf-security then we can achieve beyond birthday prf-security for the composition function ${ }^{4}$. However, obtaining a double-block beyond birthday secure prf based on a (single-keyed) block cipher would not be easy and efficient. One may try some variants of 6 rounds Luby-Rackoff [19] or Benes-Butterfly construction [21]. However, we do not know any such single key efficient construction.

### 5.1 Hash-Then-Sum Construction

In this paper, we consider a special and very simple form of $g$ function, namely the sum function over two blocks, which is considered in $[2,14]$. We define

$$
\operatorname{sum}^{\pi_{1}}(x, y)=\pi_{1}(x) \oplus \pi_{1}(y), \text { and } \operatorname{sum}^{\pi_{1}, \pi_{2}}(x, y)=\pi_{1}(x) \oplus \pi_{2}(y)
$$

where $\pi_{1}$ and $\pi_{2}$ are two independent $n$-bit functions (possibly permutations). Given a double-block construction $\mathcal{H}_{K}$, let's consider the following three composition rules depending on key reuse.

1. three-key construction $\mathcal{C}_{3}^{K, \pi_{1}, \pi_{2}}:=\operatorname{sum}^{\pi_{1}, \pi_{2}} \circ \mathcal{H}_{K}$.
2. two-key construction $\mathcal{C}_{2}^{K, \pi_{1}}:=\operatorname{sum}^{\pi_{1}} \circ \mathcal{H}_{K}$.
3. one-key construction $\mathcal{C}_{1}^{\pi}:=\operatorname{sum}^{\pi} \circ \mathcal{H}^{\pi}$.

Note that we can not apply the above composition result as the sum construction is clearly not a prf over two blocks. So we need a different type of composition result for sum-based construction. In [6], it has been proved that
 The same can be proved for PRF security instead of unforgeable.

[^1]

Fig. 5.1. Hash-then-Sum construction

### 5.1.1 Hash-then-sum based on PRF.

 isfies the following:

$$
\mathbf{A d v}_{\mathcal{C}_{3}}^{\mathrm{prf}}(t, q, \ell) \leq \mathbf{A d v}_{\mathcal{H}}^{\mathrm{cf}}(q, \ell)+2 \mathbf{A d v}_{f}^{\mathrm{prf}}\left(t^{\prime}, q, \ell\right)
$$

Proof. Fix a cover-free tuple $\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}$. We denote the event

$$
E(\sigma, \theta) \equiv\left(\left(\mathcal{H}_{K}\left(m_{i}\right)\right)_{i \in[q]}=\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}\right) .
$$

Therefore,

$$
E(\sigma, \theta) \equiv\left(\left(\mathcal{H}_{K}\left(m_{i}\right)\right)_{i \in[q]}=\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}\right)
$$

Therefore,

$$
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{3}}{\mapsto} t \mid E\right]=\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{3}}{\mapsto} t\right]=\operatorname{Pr}\left[\Gamma_{1}\left(\sigma_{i}\right) \oplus \Gamma_{2}\left(\theta_{i}\right)=t_{i}, \forall i\right]=2^{-n q} .
$$

The first equation follows from the argument that the randomness for $\mathcal{H}$ is independent of $\Gamma_{1}$ 's and $\Gamma_{2}$ 's. The last equality follows from the following argument. Let $\psi_{i}$ denote the one of the fresh blocks from $\sigma_{i}$ and $\theta_{i}$ and $\psi_{i}^{\prime}$ denotes the other. Then, by conditioning on the output of $\psi_{i}^{\prime}$ 's the above probability becomes the interpolation probability of a uniform random function for $q$ distinct inputs which equals to $2^{-n q}$. As the conditional probability is same for all condition events, the unconditional probability is also equal to $2^{-n q}$. Therefore,

$$
\begin{aligned}
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{3}}{\mapsto} t\right] & =\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{3}}{\mapsto} t \mid E\right] \times \operatorname{Pr}[E] \\
& \leq \frac{(1-\epsilon)}{2^{n q}}
\end{aligned}
$$

where $\epsilon:=\operatorname{Pr}\left[E^{c}\right]$.

Remark 1 The above three-key construction is a potential candidate for having beyond birthday security. Note that from Definition2, $\mathbf{A d v}_{\mathcal{H}}^{\mathrm{cf}}(q, \ell) \leq q^{3} \mathbf{A d v}_{\mathcal{H}}^{\mathrm{cf}}(3, \ell)$. So, for any three messages $m_{1}, m_{2}, m_{3}$ with $m_{1} \neq m_{2}, m_{3}$, if

$$
\operatorname{Pr}\left[\Sigma_{1}=\Sigma_{2}, \Theta_{1}=\Theta_{3}\right]=\mathcal{O}\left(\ell^{c} 2^{-2 n}\right)
$$

for some small constant $c$ then we have the beyond birthday security for small $\ell$. Intuitively, the event $\Sigma_{1}=\Sigma_{2}, \Theta_{1}=\Theta_{3}$ deals two (possibly linear independent) equations and it is feasible to have such a bound.

### 5.1.2 Hash-then-Sum based on Pseudorandom Permutation.

Abstraction of PMAC+, 3kf9 PMAC_Plus [27] and 3kf9 [28] are blockcipher (assumed to be a pseudorandom permutation) based sum constructions. These are three-key construction like $\mathcal{C}_{3}$. After modeling a blockcipher to be a prf, one can apply the above Lemma 4 . However, block cipher can ensure prf with a maximum birthday bound security. So we need to treat it differently to have beyond birthday analysis. The designers of PMAC_Plus and 3kf9 have proved the security for these individual constructions. Here, we abstract their analysis and provide a generic composition results. In the following, let $\Pi, \Pi_{1}, \Pi_{2}$ be random permutations over the domain $\{0,1\}^{n}$ and range $\{0,1\}^{n}$. We state the results for the constructions using uniform random permutations instead of pseudorandom permutation as the standard reduction can be applied for the later constructions. As $\mathcal{H}_{K}$ is a double block construction, we write $\mathcal{H}_{K}=\left(\mathcal{H}_{K, 1}, \mathcal{H}_{K, 2}\right)$ where $\mathcal{H}_{K, 1}, \mathcal{H}_{K, 2}$ are single block functions.

Theorem 6. Let $\mathcal{H}_{K}$ be a $\left(q, \epsilon_{c f}\right)$-cover-free function and for all $i=1,2, \mathcal{H}_{K, i}$ are $\epsilon_{\text {univ-universal hash functions. Then, the following holds. }}^{\text {fol }}$

1. $\mathcal{C}_{3}:=\operatorname{sum}^{\Pi_{1}, \Pi_{2}} \circ \mathcal{H}_{K}$ is $(q, \ell, \epsilon)$-prf where

$$
\epsilon=\epsilon_{c f}+\left(q+\frac{q^{2}}{2^{n}}\right) \epsilon_{\text {univ }}+\frac{6 q^{3} \ell^{3}}{2^{2 n}}
$$

2. $\mathcal{C}_{2}:=\operatorname{sum}^{\Pi} \circ \mathcal{H}_{K}$ is $(q, \ell, \epsilon)$-prf where

$$
\epsilon=\epsilon_{\mathrm{cf}}+\left(2 q+\frac{2 q^{2}}{2^{n}}\right) \epsilon_{\text {univ }}+\frac{6 q^{3} \ell^{3}}{2^{2 n}}
$$

Proof. The proofs for both constructions are similar except that we have to analyze sum of two independent or dependent uniform random random permutations. As the later involves more dependency, we only prove for $\mathcal{C}_{2}$. We provide the proof by using coefficient H -technique for which it would be sufficient to obtain a lower bound of interpolation probability.

Informally, given that we obtain cover-free outputs $\left(\sigma_{i}, \theta_{i}\right)_{i}$ from $\mathcal{H}$, for all $i$ at least one block is fresh. If both are fresh then we call $i$ free. For all non-free indices $i$, exactly one, denoted $\psi_{i}$, of $\sigma_{i}$ and $\theta_{i}$ is not fresh and the other denoted by $\psi_{i}^{\prime}$, is fresh. We sample the output $\Pi\left(\psi_{i}^{\prime}\right)$ which will be forced as the sum of these values
are fixed. Note that in the interpolation probability calculation, we fix some values for sum beforehand. Now, this will have high interpolation probability due to low collision probability of $\mathcal{H}_{K, i}$ 's and independence of sampling $\Pi$. In this way, we obtain high interpolation probability except for free $i$. Now we can apply sum of a uniform random permutation sampled from a restricted class of permutation to complete the interpolation probability for free indices.
Bad view. A tuple $t:=\left(t_{1}, \ldots, t_{q}\right)$ is said to have a $r$-collision if there exists an $r$-set $I$ such that $t_{i}=t_{j}$ for all $i, j \in I$. Let

$$
\begin{equation*}
\mathcal{V}_{\text {bad }}=\left\{t: \exists i, \quad t_{i}=0\right\} \cup\{t: t \text { has 3-collision }\} . \tag{3}
\end{equation*}
$$

For a random function $\Gamma$ and for any adversary $\mathcal{A}$,

$$
\begin{equation*}
\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\Gamma}\right) \in \mathcal{V}_{b a d}\right] \leq \frac{q}{2^{n}}+\frac{q^{3}}{2^{2 n}} \tag{4}
\end{equation*}
$$

Now we fix any $t \notin \mathcal{V}_{\text {bad }}$ and a $q$-tuple $m$ of distinct messages. We write $\mathcal{H}_{K}\left(m_{i}\right)=\left(\Sigma_{i}, \Theta_{i}\right), 1 \leq i \leq q$. Let $\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}$ be any tuple.
For any $i$ exactly one of the these will happen:
(i) $i$ is free
(ii) $\sigma_{i}$ is fresh and $\theta_{i}$ is not
(iii) $\theta_{i}$ is fresh and $\sigma_{i}$ is not
(iv) both $\sigma_{i}$ and $\theta_{i}$ are not fresh.

Now let $I_{\Sigma}=\left\{i: \sigma_{i}\right.$ is not fresh $\}$ and similarly we define $I_{\Theta}$. We define

$$
\left(\psi_{i}, \psi_{i}^{\prime}\right)=\left\{\begin{array}{l}
\left(\sigma_{i}, \theta_{i}\right), \text { if } i \in I_{\Sigma} \\
\left(\theta_{i}, \sigma_{i}\right), \text { if } i \in I_{\Theta}
\end{array}\right.
$$

Note that $\psi_{i}^{\prime}$ 's are always fresh and $\psi_{i}$ 's are not. We write $I=I_{\Sigma} \cup I_{\Theta}$.
We call a tuple $\left(\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]},\left(\psi_{j}, w_{j}\right)_{j \in I}\right)$ good w.r.t. $t$ if all of the followings happen:

1. $E_{1} \equiv:\left(\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}\right.$ is a cover-free tuple,
2. $E_{2} \equiv$ : whenever $t_{i}=t_{j}, \sigma_{i} \neq \sigma_{j}$ and $\theta_{i} \neq \theta_{j}$,
3. $E_{3} \equiv$ : for $w_{i}^{\prime}=w_{i}+t_{i}, i \in I$, the tuple $w_{I}^{\prime} \in$ dist and
4. $E_{4} \equiv: w_{I}^{\prime} \cap w_{I}=\phi$.

Note that, $w_{j}$ is the $\Pi\left(\psi_{j}\right)$.
Note, due to the choice of the $q$-tuple $t$, at most for $q / 2$ pairs $(i, j), t_{i}=t_{j}$ can happen.
Interpolation probability for good tuple. Let us fix a good tuple as defined above. We denote the event

$$
E(\sigma, \theta, w) \equiv\left(\left(\mathcal{H}_{K}\left(m_{i}\right)\right)_{i \in[q]}=\left(\sigma_{i}, \theta_{i}\right)_{i \in[q]}, \Pi\left(\psi_{j}\right)=w_{j} \forall j \in I\right)
$$

It is easy to see that given $E$ the interpolation event $m_{I} \xrightarrow{\mathcal{C}_{2}} t_{I}$ is same as $\psi_{I}^{\prime} \stackrel{\Pi}{\longmapsto} w_{I}^{\prime}$. Also, observe that, $\psi_{I}^{\prime} \in \operatorname{dist}_{s}$ and $\psi \cap \psi^{\prime}=\phi$ where $s=|I|$. Due to
the definition of good tuple, $w_{I}^{\prime} \in \operatorname{dist}_{s}, w_{I}^{\prime} \cap w_{I}=\phi$. Whenever $t_{i}=t_{j}$, we have $w_{i} \neq w_{j}$ as $w_{i}^{\prime} \neq w_{j}^{\prime}$. At the same time, by definition of good tuple we know that $\sigma_{i} \neq \sigma_{j}$ and $\theta_{i} \neq \theta_{j}$. So, $\left(\psi_{I}^{\prime}, \psi_{I}\right) \longleftrightarrow\left(w_{I}^{\prime}, w_{I}\right)$.
Combining all these, we have

$$
\begin{aligned}
\operatorname{Pr}\left[m_{I} \stackrel{\mathcal{C}_{2}}{\longrightarrow} t_{I} \mid E\right] & =\operatorname{Pr}\left[\psi_{I}^{\prime} \stackrel{\Pi}{\longmapsto} w_{I}^{\prime} \mid E\right] \\
& =\operatorname{Pr}\left[\psi_{I}^{\prime} \stackrel{\Pi}{\longleftrightarrow} w_{I}^{\prime} \mid \psi_{I} \stackrel{\Pi}{\longleftrightarrow} w_{I}\right] \quad \text { (As } K \text { and } \Pi \text { are independent) } \\
& \geq \frac{1}{2^{n s}} \quad\left(\operatorname{As}\left(\psi_{I}^{\prime}, \psi_{I}\right) \longleftrightarrow\left(w_{I}^{\prime}, w_{I}\right), \psi_{I}^{\prime} \cap \psi_{I}=\phi \text { and } \psi_{I}^{\prime} \in \operatorname{dist}_{s}\right)
\end{aligned}
$$

Using the above result, we find the following conditional probability

$$
\begin{aligned}
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{2}}{\longmapsto} t \mid E\right] & =\operatorname{Pr}\left[m_{I^{c}} \stackrel{\mathcal{C}_{2}}{\longmapsto} t_{I^{c}} \mid E \wedge m_{I} \stackrel{\mathcal{C}_{2}}{\longmapsto} t_{I}\right] \times \operatorname{Pr}\left[m_{I} \stackrel{\mathcal{C}_{2}}{\longmapsto} t_{I} \mid E\right] \\
& \geq \operatorname{Pr}\left[\left(\sigma_{i}, \theta_{i}\right)_{i \in I^{c^{\prime}}} \stackrel{\text { sum }}{\longmapsto} t_{I^{c}} \mid\left(\psi_{I}, \psi_{I}^{\prime}\right) \stackrel{\Pi}{\longmapsto}\left(w_{I}, w_{I}^{\prime}\right)\right] \times \frac{1}{2^{n s}} \\
& \geq \frac{\left(1-6 s^{3} / 2^{2 n}\right)}{2^{n q}} \quad[\text { From (b) of Corollary 2] }
\end{aligned}
$$

Now, we find our desired interpolation probability as we sum over all good tuples:

$$
\begin{aligned}
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{2}}{\longmapsto} t\right] & \geq \sum_{E} \operatorname{Pr}\left[m \stackrel{\mathcal{C}_{2}}{\longmapsto} t \mid E\right] \times \operatorname{Pr}[E] \\
& \geq \frac{\left(1-6 s^{3} / 2^{2 n}\right)}{2^{n q}} \times(1-\epsilon)
\end{aligned}
$$

where $\epsilon=\operatorname{Pr}\left[\left(\Sigma_{i}, \Theta_{i}\right)_{i \in[q]},\left(\Psi_{i}, \Pi\left(\Psi_{i}\right)\right)_{i \in I}\right.$ is not good $]$.
Bounding $\epsilon$. By using the definition of good tuple and using the union bound, we have $\epsilon \leq \epsilon_{1}+\epsilon_{2}+\epsilon_{3}+\epsilon_{4}$ where $\epsilon_{i}=\operatorname{Pr}\left[E_{i}^{c}\right], 1 \leq i \leq 4$. Now we bound each $\epsilon_{i}$ as follows:
(a) $\epsilon_{1}=\operatorname{Pr}\left[\left(\Sigma_{i}, \Theta_{i}\right)_{i}\right.$ is not cover-free $] \leq \epsilon_{\mathrm{cf}}$.
(b) $\epsilon_{2}=\sum_{i \neq j: t_{i}=t_{j}}\left(\operatorname{Pr}\left[\Sigma_{i}=\Sigma_{j}\right]+\operatorname{Pr}\left[\Theta_{i}=\Theta_{j}\right]\right) \leq 2 q \epsilon_{\text {univ }}$.
(c) $\epsilon_{3}=\operatorname{Pr}\left[w_{I}^{\prime} \in \mathrm{dist}\right] \leq \frac{q^{2}}{2^{n}} \epsilon_{\text {univ }}$. The proof is given below:

$$
\begin{aligned}
& \epsilon_{3}=\sum_{i \neq j: t_{i} \neq t_{j}} \operatorname{Pr}\left[i, j \in I, \Pi\left(\Psi_{i}\right) \oplus \Pi\left(\Psi_{j}\right)=t_{i} \oplus t_{j}\right] \\
& \leq \sum_{i, j, k, \psi_{i}, \psi_{j}: i \neq j, t_{i} \neq t_{j}} \operatorname{Pr}\left[\Pi\left(\Psi_{i}\right) \oplus \Pi\left(\Psi_{j}\right)=t_{i} \oplus t_{j} \mid \Psi_{i}=\Psi_{k}=\psi_{i}, \Psi_{j}=\psi_{j}\right] \\
& \times \operatorname{Pr}\left[\Psi_{i}=\Psi_{k}=\psi_{i}, \Psi_{j}=\psi_{j}\right] \\
& \leq \sum_{i, j, k, \psi_{i}, \psi_{j}: i \neq j, t_{i} \neq t_{j}} \operatorname{Pr}\left[\Pi\left(\psi_{i}\right) \oplus \Pi\left(\psi_{j}\right)=t_{i} \oplus t_{j}\right] \\
& \times \operatorname{Pr}\left[\Psi_{i}=\Psi_{k}=\psi_{i}, \Psi_{j}=\psi_{j}\right] \\
& \leq \sum_{i, k} \frac{1}{2^{n}-1} \times \operatorname{Pr}\left[\Psi_{i}=\Psi_{k}\right]
\end{aligned}
$$

The last two inequalities follows from the two fact: (i) $K$ is independent of $\Pi$ and (ii) for any $a, b, \operatorname{Pr}[\Pi(a) \oplus \Pi(b)=c] \leq 1 /\left(2^{n}-1\right)$
(d) $\epsilon_{4}=\operatorname{Pr}\left[w_{I}^{\prime} \cap w_{I}=\phi\right]=\sum_{i \neq j: t_{i} \neq t_{j}} \operatorname{Pr}\left[i, j \in I, \Pi\left(\Psi_{i}\right) \oplus \Pi\left(\Psi_{j}\right)=t_{i}\right] \leq \frac{q^{2}}{2^{n}} \epsilon_{\text {univ }}$.

This proof is identical to case (c).
Adding these four error terms, we obtain an upper bound of $\epsilon$. By using coefficient $H$-technique, our result follows.

### 5.2 PRF-security of Single Key Hash-then-Sum Construction

In this paper, we show a prf-security bound for one-key hash-then-sum constructions $\mathcal{C}_{1}:=\operatorname{sum}^{\Pi} \circ \mathcal{H}^{\pi}$. Note that the hash function is also permutation based and uses same permutation $\Pi$ used in the outer layer sum function. The PRF security analysis is similar to that of Theorem 6 . However, it requires to handle more bad cases. Now we first develop the basic notations and definitions similar to the two-key and three-key constructions.

Given any permutation $\pi$, let $\tau\left(\mathcal{H}(m) \rightarrow_{q} \pi\right)=(x, y)$ (the pair of inputs and outputs of $\pi$ during the computations of $\mathcal{H}^{\pi}\left(m_{i}\right)=\left(\sigma_{i}, \theta_{i}\right)$ for all $i \in[q]$. We also write $x=\left(x_{i, j}: i \in[q], j \in\left[L_{i}\right]\right)$ and similarly $y$ for the same index set. Note that $\left(\sigma_{i}, \theta_{i}\right)_{i}$ is uniquely determined by $(x, y)$.

Definition 4. For any $i$, we say that $\sigma_{i}$ is $x$-fresh if it is not same as $\sigma_{j}$ for some $j \neq i$ or $x_{k}$ for any $k$. Similarly, we define for $x$-freshness of $\theta_{i}$. We say that a tuple $\left(\sigma_{i}, \theta_{i}\right)_{i}$ is $x$-cover-free (or $(x, y)$ is extended-cover-free) if for all $i$, either $\sigma_{i}$ or $\theta_{i}$ (or both) is $x$-fresh. If both $\sigma_{i}$ and $\theta_{i}$ are $x$-fresh we call $i$ to be free.

We denote $I_{\Sigma}=\left\{i: \sigma_{i}\right.$ is not $x$-fresh $\}$ and similarly $I_{\Theta}$ and let $I=I_{\Sigma} \cup I_{\Theta}$. For all $i \in I_{\Sigma}$, we define $\left(\psi_{i}, \psi_{i}^{\prime}\right)=\left(\sigma_{i}, \theta_{i}\right)$ and similarly, for all $i \in I_{\Theta}$, we define $\left(\psi_{i}^{\prime}, \psi_{i}\right)=\left(\sigma_{i}, \theta_{i}\right)$ and so $\psi_{i}^{\prime}$ 's are always non-fresh and $\psi_{i}^{\prime}$ 's are fresh. We say that $\psi_{i}$ is old if there exists $x_{j}$ such that $\psi_{i}=x_{j}$, otherwise $\psi_{i}$ is called new. We define $I_{\text {old }}=\left\{i: \psi_{i}\right.$ is old $\}$ and similarly $I_{\text {new }}=\left\{i: \psi_{i}\right.$ is new $\}$. Let $I=I_{\text {old }} \cup I_{\text {new }}$.

Definition 5. We say that a tuple $\left((x, y), w_{I_{\text {new }}}\right)$ good if followings happen:

1. $E_{1} \equiv(x, y)$ is extended-cover-free,
2. $E_{2} \equiv$ whenever $t_{i}=t_{j}, \sigma_{i} \neq \sigma_{j}$ and $\theta_{i} \neq \theta_{j}$,
3. $E_{3} \equiv\left(x, \psi_{I}, \psi_{I}^{\prime}\right) \longleftrightarrow\left(y, w_{I}, w_{I}^{\prime}\right)$ where $w_{i}=y_{j, a}$ for all $i \in I_{\text {old }}$ with $\psi_{i}=$ $x_{j, a}$ and $w_{i}^{\prime}=w_{i}+t_{i}, \forall i \in I$.

By definition of $I$ and $\left(\psi_{I}, \psi_{I}^{\prime}\right)$ we have (i) $\psi_{I}^{\prime} \in$ dist and (ii) $\psi_{I}^{\prime} \cap\left(x, \psi_{I}\right)=\phi$. Thus, the event $E_{3}^{c}$ (in presence of $E_{1}$ and $E_{2}$ ) is equivalent to at least one of the following events happen:

1. $w_{i} \oplus w_{j}=t_{i} \oplus t_{j}$ for some $i, j \in I$ such that $t_{i} \neq t_{j}$.
2. $w_{i} \oplus t_{i}=y_{j, a}$ or $=w_{k}$ for some $k \in I$,
3. $w_{i}=y_{j, a}$ for some $i \in I_{\text {new }}$,
4. $\left(\psi_{I_{\text {new }}}, w_{I_{\text {new }}}\right)$ is permutation compatible.

Note that the 4 th bad equations can be easily avoided by choosing $w_{I_{\text {new }}}$ such that $\left(\psi_{I_{\text {new }}}, w_{I_{\text {new }}}\right)$ is a permutation compatible. Now we identify and explicitly list down all the bad equations for which the tuple $\left((x, y), w_{I_{\text {new }}}\right)$ is not good, in Table 5.2.

| Fully Covered | $\begin{aligned} & \left(L_{11}\right) \Sigma_{i}=\Sigma_{j}, \Theta_{i}=\Theta_{k} \\ & \left(L_{12}\right) \\ & \Sigma_{i}=X_{j, a}, \Theta_{i}=\Theta_{k} \\ & \left(L_{13}\right) \\ & \Sigma_{i}=\Sigma_{j}, \Theta_{i}=X_{k, b} \\ & \left(L_{14}\right) \\ & \Sigma_{i}=X_{j, a}, \Theta_{i}=X_{k, b} \\ & \hline \end{aligned}$ |
| :---: | :---: |
| ( $X, Y$ )-Pseudo Cover-1 | $\begin{aligned} & \left(L_{21}\right) \Sigma_{i}=X_{j, a}, Y_{j, a} \oplus t_{i}=Y_{k, s} \\ & \left(L_{22}\right) \Theta_{i}=X_{j, a}, Y_{j, a} \oplus t_{i}=Y_{k, s} \\ & \hline \end{aligned}$ |
| ( $X, Y$ )-Pseudo Cover-2 | $\left(L_{23}\right) \Sigma_{i}=X_{k, a}, \Sigma_{j}=X_{l, b}, Y_{k, a} \oplus Y_{l, b}=t_{i} \oplus t_{j}$ $\left(L_{24}\right) \Theta_{i}=X_{k, a}, \Theta_{j}=X_{l, b}, Y_{k, a} \oplus Y_{l, b}=t_{i} \oplus t_{j}$ $\left(L_{25}\right) \Sigma_{i}=X_{k, a}, \Theta_{j}=X_{l, b}, Y_{k, a} \oplus Y_{l, b}=t_{i} \oplus t_{j}$ |
| $\left(X, Y, w_{I_{\text {new }}}\right)$ <br> Pseudo Covered | $\begin{aligned} & \left(L_{31}\right) \Sigma_{i}=X_{j, a}, Y_{j, a} \oplus t_{i}=w_{k, s} \\ & \left(L_{32}\right) \Theta_{i}=X_{j, a}, Y_{j, a} \oplus t_{i}=w_{k, s} \\ & \left(L_{33}\right) \Sigma_{i}=\Sigma_{j}, w_{i}+t_{i}=w_{j}+t_{j} \\ & \left(L_{34}\right) \Sigma_{i}=\Sigma_{j}, w_{i}+t_{i}=w_{j} \\ & \left(L_{35}\right) \Theta_{j}=X_{l, b}, Y_{k, a} \oplus Y_{l, b}=t_{i} \oplus t_{j} \end{aligned}$ |

Table 1. Table representing bad equations for fully covered, pseudo-covered cases.

Definition 6. 1. A construction $\mathcal{H}^{\Pi}$ is called $(q, \ell, \epsilon)$-extended-cover-free if for all $q$-tuple $m$ of distinct messages of size at most $\ell, \operatorname{Pr}_{\Pi}[\exists$ fully covered $i] \leq$ $\epsilon$.
2. It is called $(q, \ell, \epsilon)$-pseudo-cover-free $w . r . t . t$ if for all $q$-tuple $m$ of distinct messages of size at most $\ell$, if $\operatorname{Pr}_{\Pi}[\exists i: i$ is $(X, Y)$ pseudo-covered $] \leq \epsilon:=$ $\epsilon_{1}+\epsilon_{2}$ where $\epsilon_{1}:=\operatorname{Pr}_{\Pi}[\exists i: i$ is $(X, Y)$-pseudo-cover- 1$]$ and $\epsilon_{2}:=\operatorname{Pr}_{\Pi}[\exists i:$ $i$ is $(X, Y)$-pseudo cover-2].
3. It is called $\epsilon$-extended universal if $\mathcal{H}_{i}^{\Pi}$ 's are $\epsilon$ universal and for all pairs $m=\left(m_{1}, m_{2}\right)$ of distinct messages $\operatorname{Pr}_{\Pi}\left[\Sigma_{1}=X_{i, j}\right], \operatorname{Pr}\left[\Theta_{1}=X_{i, j}\right] \leq \epsilon$ for all $i=1,2$ and $j \in\left[L_{i}\right]$.

Theorem 7. If $\mathcal{H}$ is block-separated, $\left(q, \ell, \epsilon_{e c f}\right)$-extended-cover-free, $\left(q, \ell, \epsilon_{p c f}\right)$ -pseudo-cover-free for a $q$-tuple $t$ and $\epsilon_{\text {euniv-extended }}$ universal then $\mathcal{C}_{1}:=$ sum $^{\Pi} \circ$ $\mathcal{H}^{\Pi}$ is $(q, \epsilon)$-prf where

$$
\epsilon=\epsilon_{e c f}+\epsilon_{p c f}+2 q \epsilon_{e u n i v}++\frac{18 s^{3}}{2^{2 n}}
$$

Note that we should expect $O\left(s^{3} / 2^{2 n}\right)$ errors for $\epsilon_{e c f}$ and $\epsilon_{p c f}$ as it deals two (apparently) non-trivial equations. If so, then only we can claim beyond birthday security for the construction.

Proof. Bad view: A tuple $t:=\left(t_{1}, \ldots, t_{q}\right)$ is said to have $r$-collision if there exists an $r$-set $I$ such that $t_{i}=t_{j}$ for all $i, j \in I$. Let

$$
\mathcal{V}_{\text {bad }}=\left\{t: \exists i, \quad t_{i}=0\right\} \cup\{t: t \text { has 3-collision }\} .
$$

For a random function $\Gamma$ and for any adversary $\mathcal{A}$,

$$
\operatorname{Pr}\left[\tau\left(\mathcal{A}^{\Gamma}\right) \in \mathcal{V}_{b a d}\right] \leq \frac{q}{2^{n}}+\frac{q^{3}}{2^{2 n}}
$$

Now we fix any $t \notin \mathcal{V}_{b a d}$ and a $q$-tuple $m$ of distinct messages.
Interpolation probability for good tuple. Let us fix a good tuple $\left((x, y), w_{I_{\text {new }}}\right)$ as defined in Definition 5. We denote the event

$$
E(x, y, w) \equiv\left(x \stackrel{\Pi}{\longmapsto} y, \Pi\left(\Psi_{i}\right)=w_{i} \forall i \in I_{n e w}\right)
$$

. It is easy to see that given $E$ the interpolation event $m_{I} \stackrel{\mathcal{C}_{1}}{\longmapsto} t_{I}$ is same as $\psi_{I}^{\prime} \stackrel{\Pi}{\longmapsto} w_{I}^{\prime}$. Also observe that $\psi_{I}^{\prime} \in$ dist and (ii) $\psi_{I}^{\prime} \cap\left(x, \psi_{I}\right)=\phi$ where $s=|I|$. Due to the definition of good tuple $w_{I^{\prime}} \in$ dist and $w_{I^{\prime}} \cap w_{I}=\phi$. Moreover $(x, y)$ is permutation computable for $\Pi$. Therefore, $\left(x, \psi_{I}, \psi_{I}^{\prime}\right) \longleftrightarrow\left(y, w_{I}, w_{I}^{\prime}\right)$, where $\psi_{I}^{\prime}$ is element-wise distinct and distinct from other inputs.

Combining all these, we have

$$
\begin{aligned}
\operatorname{Pr}\left[m_{I} \stackrel{\mathcal{C}_{1}}{\longmapsto} t_{I} \mid E\right] & =\operatorname{Pr}\left[\psi_{I}^{\prime} \stackrel{\Pi}{\longmapsto} w_{I}^{\prime} \mid E\right] \\
& =\operatorname{Pr}\left[\psi_{I}^{\prime} \stackrel{\Pi}{\longmapsto} w_{I}^{\prime} \mid\left(x, \psi_{I}\right) \stackrel{\Pi}{\longmapsto}\left(y, w_{I}\right)\right] . \\
& \geq \frac{1}{2^{n s}} \mathrm{As},\left(\psi_{I}^{\prime}, \psi_{I}\right) \longleftrightarrow\left(w_{I}^{\prime}, w_{I}\right), \psi_{I}^{\prime} \cap\left(x, \psi_{I}\right)=\phi, \psi_{I}^{\prime} \in \operatorname{dist}
\end{aligned}
$$

Using the above result, we find the following conditional probability

$$
\begin{aligned}
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{1}}{\longrightarrow} t \mid E\right] & =\operatorname{Pr}\left[m_{I^{c}} \stackrel{\mathcal{C}_{2}}{\longmapsto} t_{I^{c}} \mid E \wedge m_{I} \stackrel{\mathcal{C}_{1}}{\longmapsto} t_{I}\right] \times \operatorname{Pr}\left[m_{I} \stackrel{\mathcal{C}_{1}}{\longmapsto} t_{I} \mid E\right] \\
& \geq \operatorname{Pr}\left[\left(\sigma_{i}, \theta_{i}\right)_{i \in I^{c}} \stackrel{\text { sum }}{\longmapsto} t_{I^{c}} \mid\left(x, \psi_{I}, \psi_{I}^{\prime}\right) \stackrel{\Pi}{\longmapsto}\left(y, w_{I}, w_{I}^{\prime}\right)\right] \times \frac{1}{2^{n s}} \\
& \geq 2^{-n q} \times\left(1-18 s^{3} / 2^{2 n} .\right) \quad[\text { From Corollary 2]. }
\end{aligned}
$$

Now, we find our desired interpolation probability as we sum over all good tuples:

$$
\begin{aligned}
\operatorname{Pr}\left[m \stackrel{\mathcal{C}_{1}}{\longmapsto} t\right] & \geq \sum_{E} \operatorname{Pr}\left[m \stackrel{\mathcal{C}_{1}}{\longmapsto} t \mid E\right] \times \operatorname{Pr}[E] \\
& \geq \frac{\left(1-18 s^{3} / 2^{2 n}\right)}{2^{n q}} \times(1-\epsilon)
\end{aligned}
$$

where $\epsilon=\operatorname{Pr}\left[(X, Y)_{i \in[q]},\left(\Pi\left(\Psi_{j}\right)\right)_{j \in I}\right.$ is not good $]$.
Bounding $\epsilon$. By using the definition of good tuple and using the union bound, we have $\epsilon \leq \epsilon_{1}+\epsilon_{2}+\epsilon_{3}$ where $\epsilon_{i}=\operatorname{Pr}\left[E_{i}^{c}\right], 1 \leq i \leq 3$. Now we bound each $\epsilon_{i}$ as
follows
(a) $\epsilon_{1}=\operatorname{Pr}[(x, y)$ is not extended-cover-free $] \leq \epsilon_{\text {ecf }}$.
(b) $\epsilon_{2}=\sum_{i \neq j: t_{i}=t_{j}}\left(\operatorname{Pr}\left[\Sigma_{i}=\Sigma_{j}\right]+\operatorname{Pr}\left[\Theta_{i}=\Theta_{j}\right]\right) \leq 2 q \epsilon_{\text {univ }}$.
(c) $\epsilon_{3}=\epsilon_{\mathrm{pcf}}$ as the event $E_{3}^{c}$ in presence of $E_{1}$ and $E_{2}$ is same as violating pseudo-cover-free.
Summing these four error terms, we obtain an upper bound of $\epsilon$. The rest follows by using coefficient $H$-technique.

## 6 A Generic PRF Bound using Rank and Accident

### 6.1 Some Notes from Linear Algebra

A linear equation $L\left(X_{1}, \ldots, X_{s}\right):=L_{1} \cdot X_{1}+\cdots+L_{s} \cdot X_{s}$ over the finite field $\mathbb{F}_{2^{n}}{ }^{5}$ of size $2^{n}$ with $s$ variables can be identified as an $s$-tuple $\left(L_{1}, \ldots, L_{s}\right)$. Let $\mathcal{L}=\left\{L_{1}, \ldots, L_{q}\right\}$ be a $q$-set of linear equations with $s$-variable, then $\mathcal{L}$ can be viewed as an $q \times s$ matrix $\mathbf{L}:=\left(\left(L_{i, j}\right)\right)_{i, j}$ where $L_{i, j}$ is the $j^{\text {th }}$ coefficient of $L_{i}$. $\operatorname{rank}(\mathcal{L})$ denotes the rank of the matrix $\mathbf{L}$.
Reducing Linear Equations By Eliminating Dependent Variables. Let $L$ be a $s$-variable linear equation over $\mathbb{F}_{2^{n}}$. Then, given any equivalence relation $\sim$ over $[s]$ one can reduce the equation $L$ by eliminating dependent variables assuming that the variables induces the collision relation $\sim$. For example let $L=X_{1}+a X_{2}+X_{3}+b X_{4}+c X_{5}$ for some constant $a, b, c$ and let $\sim$ be an equivalence relation on [5] corresponding to the partition $\{\{1,3,4\},\{2,5\}\}$. If $X:=\left(X_{1}, X_{2}, X_{3}, X_{4}, X_{5}\right)$ induces $\sim$ then $X_{1}=X_{3}=X_{4}$ and $X_{2}=X_{5}$. So, by eliminating $X_{3}, X_{4}, X_{5}$, the equation $L(X)$ can be simplified to $b X_{1}+(a+c) X_{2}$. One can also eliminate $X_{1}, X_{3}, X_{5}$ and so the choice of free and determined variables are not unique. In this paper we keep the variables whose indices are minimum w.r.t. some natural order. Let $\sim$ have $c$ classes and $I=\left\{i_{1}, \ldots, i_{c}\right\}$ be the set consisting of all minimum elements from each $c$ classes. The $X_{I}$ is a tuple of free variables and the rest of the variables can be uniquely determined from $X_{I}$. After eliminating the determined variables, the simplified (also called reduced) equation would be denoted by $L^{\sim}\left(X_{I}\right)$. Note that

$$
\begin{equation*}
\text { for all } x, \quad \sim_{x}=\sim \Rightarrow L^{\sim}\left(x_{I}\right)=L(x) . \tag{5}
\end{equation*}
$$

We can also reduce when the restrictions among variables are some general linear equations instead of equality or collision relation (which is also a special form of linear equations). Let $\mathcal{R}$ be a set of linear equations over $s$-tuple of variables $X$ and $L(X)$ be the target linear equation which is going to be reduced by applying the restriction $\mathcal{R}$. We can then similarly reduce the equation $L$ by eliminating the dependent variables with free variablesof $\mathcal{R}$ after applying the linear restrictions

[^2]$\mathcal{R}$. Let $X_{I}$ be the free variables in $\mathcal{R}$ which determine the rest of the variables. ${ }^{6}$ Note that $|I|=s-\operatorname{rank}(\mathcal{R})$. Then by applying the linear dependencies of $X_{I^{c}}$ on $X_{I}$, we can reduce $L(X)$ to an equation of the form $L^{\mathcal{R}}\left(X_{I}\right)$. We similarly have
\[

$$
\begin{equation*}
\forall L^{\prime} \in \mathcal{R}, L^{\prime}(x)=0 \Rightarrow L^{\mathcal{R}}\left(x_{I}\right)=L(x) \tag{6}
\end{equation*}
$$

\]

In the last section, we have seen that one-key sum-based construction can be bounded in term of the advantages of extended-cover-free, pseudo-cover-free and universal properties of the underlying construction $\mathcal{C}^{\Pi}$. In case of an affine mode, all these advantages are probability of some affine equations over $Y$, the intermediate output tuple. Even if the equations happens to be linearly independent, we cannot have an estimate of these events (i.e extended-cover-free, pseudo-cover-free and universal) as $Y_{i}$ 's are dependent. So we need to identify a sub-tuple $Y_{I}$ which behaves "like uniform and independent random variables" and then express the linear equations in terms of $Y_{I}$. In the following subsection, we formally define what we mean by "behaves like uniform and independent".

### 6.2 Almost Independent Sampling

WR sampling is an independent sampling, but WOR is not. But they share common features in terms of conditional entropy. In particular, the conditional distribution of $i^{\text {th }}$ sample has high entropy when $i$ is not very close to total population size. We formally define it by almost-independence.

Definition 7. $\left(X_{1}, \ldots, X_{q}\right)$ is called $\epsilon$-almost-independent if for all $t_{1}, \ldots, t_{q}$, and for all $i$, the conditional probability

$$
\operatorname{Pr}\left[X_{i}=t_{i} \mid X_{1}=t_{1}, \ldots, X_{i-1}=t_{i-1}\right] \leq \epsilon
$$

If $\left(X_{1}, \ldots, X_{q}\right) \stackrel{\text { wr }}{\leftarrow} S$ then $\left(X_{1}, \ldots, X_{q}\right)$ is also $|S|^{-1}$-almost-independent. Similarly, if $\left(X_{1}, \ldots, X_{q}\right) \stackrel{\text { wor }}{\leftarrow} S$ then $\left(X_{1}, \ldots, X_{q}\right)$ is also $(|S|-q)^{-1}$-almostindependent. Now we consider a different example of almost-independent random variables obtained by conditioning WR samples.

Example 1. Suppose $\left\{\mathcal{E}_{i}(x): 1 \leq i \leq s\right\}$ is a set of $s$ affine equations ${ }^{7}$ over $G F\left(2^{n}\right)$ in $q$ variables $x_{1}, \ldots, x_{q}$. We write $\neg \mathcal{E}_{i}(x)$ to denote the affine inequation. We write the set

$$
\mathcal{E}^{\prime}=\left\{\left(x_{1}, \ldots, x_{q}\right) \in G F\left(2^{n}\right)^{q}: \forall i, \neg \mathcal{E}_{i}(x)\right\} .
$$

Let $X:=\left(X_{1}, \ldots, X_{q}\right)$ be a WR samples from $S$. Then, the conditional distribution of

$$
X \mid\left(\neg \mathcal{E}_{i}(X)\right)_{i} \text { is }\left(2^{n}-s\right)^{-1} \text {-almost-independent. }
$$

[^3]As $X$ is uniform, the conditional distribution is actually uniformly sampled from the set $\mathcal{E}^{\prime}$ and hence the conditional distribution of $X_{i}$ given $\left(X_{j}\right)_{j \neq i}$ is uniform over a set of size at least $2^{n}-s$.

Lemma 5. Let $X_{1}, \ldots, X_{q}$ is $\epsilon$-almost-independent over $G F\left(2^{n}\right)$, and let $L_{1}$, ..., $L_{r}$ be $r$ linearly independent equations with $q$ variables over the finite field $G F\left(2^{n}\right)$. Then, for any constants $c_{1}, \ldots, c_{r} \in G F\left(2^{n}\right)$, we have

$$
\begin{equation*}
\operatorname{Pr}\left[L_{i}\left(X_{1}, \ldots, X_{q}\right)=c_{i}, 1 \leq i \leq q\right] \leq \epsilon^{r} . \tag{7}
\end{equation*}
$$

Proof. By using elementary operations on the vectors $L_{i}$ 's we can equivalently express the set of equations as $L_{i}^{\prime}\left(X_{1}, \ldots, X_{a_{i}}\right)=c_{i}^{\prime}, 1 \leq i \leq r$. Now, note that $X_{i}$ is almost independent conditioned on $X_{1}, \ldots, X_{i-1}$ with probability at most $\epsilon$. As there are $r$ many linearly independent equations we can find $r$ many such $X_{i}^{\prime} s$. Thus the result holds.

Remark 2 Almost-independent is important for bounding the set of linearly independent equations. In general, we can not bound it..
6.2.1 Conditional WOR Sampling Now we consider a variant of WOR sampling, called conditional WOR sampling. This sampling scheme is motivated from the affine mode. More precisely, during the computation of permutation based affine mode, the intermediate outputs forms a conditional WOR sample. Informally, depending on the previous sample values, a conditional WOR sampling scheme either makes a fresh WOR sample or it choose one of the specific previous values. Clearly, it can not be almost-independent as the sample values can be same as the previous values. Later we identify a (random) subset of the sample which would constitute an almost independent random variables.

Let $A_{i}$ be an affine equation over $G F\left(2^{n}\right)$ with $i-1$ variables, $1 \leq i \leq \sigma^{\prime}$. The samples $Y=\left(Y_{1}, \ldots, Y_{L}\right)$ is defined as follows.

1. For $i=1$ to $L$, we define $X_{i}$ and $Y_{i}$ recursively as follows:

- $X_{i}=A_{i}\left(Y_{1}, \ldots, Y_{i-1}\right)$ and
- $Y_{i}= \begin{cases}Y_{j} & \text { if for some } j<i, X_{i}=X_{j} ; \\ \stackrel{\$}{\leftarrow}\{0,1\}^{n} \backslash\left\{Y_{j}: 1 \leq j<i\right\} & \text { otherwise. }\end{cases}$

Definitely $Y_{i}$ 's are not almost-independent as $Y_{i}=Y_{j}$ for some conditional choices of $Y_{1}, \ldots, Y_{i-1}$. So we now identify a set of (random) indices $I$ for which $Y_{i}$ 's behave almost-independently for all $i \in I$. But, this $I$ is a random set and so we will consider the conditional distribution of $Y_{I}:=\left(Y_{i}\right)_{i \in I}$ given $I$ (more precisely given an equivalence relation $\sim$ which uniquely determines $I$ ). Then, this conditional distribution would behave almost-independently. The details are given below.

Definition 8. Let $Y=\left(Y_{1}, \ldots, Y_{L}\right)$ be an $A$-conditional WOR L-sample. We define an (induced) equivalence relation $\sim_{Y}$ on $[L]$ as $i \sim j$ if and only if $A_{i}(Y)=$ $A_{j}(Y)$ (and hence $Y_{i}=Y_{j}$. We say that an equivalence relation $\sim$ is realizable if $\operatorname{Pr}\left[\sim_{Y}=\sim\right]>0$.

Let $J:=\left(J_{1}, \ldots, J_{s}\right)$ be the first indices at which $X_{i}$-values (i.e., $A_{i}$ values) are fresh. In other words, these are the minimum value for the equivalence classes and hence $J_{i}$ 's are uniquely determined from $\sim$. Note that $J_{1}=1$. Moreover, $X_{i}$ can be expressed as some affine function, denoted $\bar{A}_{i}$, over $Y_{J_{i}}$ 's. In other words, $\bar{A}_{i}\left(Y_{J}\right)=A_{i}(Y)$ for all $i$. Now, consider the following set of linear equations

$$
\bar{A}_{i}\left(Y_{J_{1}}, \ldots, Y_{J_{s}}\right)=\bar{A}_{j}\left(Y_{J_{1}}, \ldots, Y_{J_{s}}\right), \quad \forall i \sim j
$$

These conditions restrict the values of $Y_{J_{i}}$ 's.
Definition 9 (accident [4, 18]). Let $\sim$ be a realizable equivalence relation. We define accident of $\sim$, denoted $\operatorname{acc}(\sim)$, the rank of the set of linear equations:

$$
\bar{A}_{i}\left(Y_{J_{1}}, \ldots, Y_{J_{s}}\right)=\bar{A}_{j}\left(Y_{J_{1}}, \ldots, Y_{J_{s}}\right), \quad \forall i \sim j
$$

Let $I \subset\left\{J_{1}, \ldots, J_{s}\right\}$ be the set of free variables of size $s-a$, which appear first, such that $Y_{I_{j}}$ 's will determine rest of the $Y$ values. We call I to be the set of free indices associated with $\sim$.

Proposition 1. Let $\sim$ be a realizable equivalence relation and let $I$ be the corresponding indices as defined above. Then, the conditional distribution of $Y_{I} \mid \sim_{Y}=\sim$ is $\left(2^{n}-L^{2}\right)^{-1}$-almost-independence.

Proof. We identify a set of inequations $\neg \mathcal{E}$ and then we show that $Y_{I} \mid \sim_{Y}=\sim$ and $U_{I} \mid \neg \mathcal{E}\left(U_{I}\right)$ have same distributions where $U_{I}$ is the WR sample. Thus from Example 1 the result follows.

### 6.3 Connection between conditional WOR sample and blockcipher based Affine Construction

Let $\mathcal{C}$ be an affine construction meaning that the intermediate inputs (the inputs of the blockcipher) is an affine function of previous intermediate outputs and message blocks. Then, all intermediate outputs of the computation of one or more messages can be viewed as a conditional WOR sampling for a suitable choices of affine functions. We can similarly define accident of a permutation for a tuple of messages.

For any pair ( $m, \pi$ ) of $q$-tuple of distinct messages and a permutation, we associate the following objects:

1. equivalence relation (which is same as the structure graph in case of CBC construction) [4] on intermediate outputs $Y$ with $s$ many classes,
2. accident $a:=\operatorname{acc}^{m}(\pi)$, representing the number of linearly independent restrictions and
3. and a set of indices $I$ of size $s-a$ such that $Y_{I}$ is $\left(2^{n}-\left(\sigma^{\prime}\right)^{2}\right)$-almost-indp where $\sigma^{\prime}$ is the total number of message blocks.

We say that a permutation is not allowed w.r.t. a $q$-tuple of distinct messages $m:=\left(m_{1}, \ldots, m_{q}\right)$, if

1. for all $i, \operatorname{acc}^{m_{i}}(\pi) \geq 1$,
2. for all $i, j, k, \operatorname{acc}^{m_{i}, m_{j}, m_{k}}(\pi) \geq 2$ and
3. for all $i, j, k, l, \operatorname{acc}^{m_{i}, m_{j}, m_{k}, m_{l}}(\pi) \geq 3$,

Lemma 6 ( [18]). For any realizable equivalence relation $\sim$ with accident $a$ $\operatorname{Pr}\left[\sim_{Y}=\sim\right] \leq \frac{1}{\left(2^{n}-L\right)^{a}}$. The number of realizable equivalence relation with accident $a$ is at most $\left(\binom{s}{2}^{a}\right)$.

We skip the proof of bounding the number of realizable equivalence relations with accident $a$. Informally, to each an $a$ accident realizable relation, we would be able to uniquely identify a basis of $a$ linear equations (there are several choices of basis, but a special way of selecting basis will ensure the uniqueness of the choice). Since each equation can be chosen at most $\binom{s}{2}$ ways, the number of ways we can choose a special basis is at most $\binom{s}{2}^{a}$.

Lemma 7. Probability that a random permutation is not allowed for a tuple of $q$ messages is at most

$$
\frac{q L^{2}}{2^{n}}+\frac{q^{2} L^{4}}{2^{2 n}}+\frac{q^{3} L^{6}}{2^{3 n}}
$$

A not allowed permutation will be treated as a bad permutation. We make our analysis for allowed permutation. Note that a permutation is allowed for a $q$-tuple of messages if and only if for all distinct $i, j, k ; \pi$ is also allowed for $\left(m_{i}, m_{j}, m_{k}\right)$.

### 6.4 PRF Bound of Single-Key Hash-then-Sum Construction through rank analysis

Lemma 8. If $\mathcal{C}$ is $(\epsilon, 3)$-extended-cover-free then $\mathcal{C}$ is $\left.\binom{q}{3} \epsilon, q\right)$-cover-free. Similarly, if $\mathcal{C}$ is $(\epsilon, 3)$-pseudo-cover-free-1 then $\mathcal{C}$ is $\left.\binom{q}{3} \epsilon, q\right)$-pseudo-cover-free- 1 . Moreover if $\mathcal{C}$ is $(\epsilon, 4)$-pseudo-cover-free-2 then $\mathcal{C}$ is $\left(\binom{q}{4} \epsilon, q\right)$-pseudo-cover-free2.

Applying this result to Theorem 7, it would be sufficient to bound, extended-cover-free for three messages and pseudo-cover-free advantages for three and four messages. However, for some constructions, we may not be able to obtain desired bound. So we need to consider allowed permutations.

Given, a set of affine equations $\mathcal{L}$ and an equivalence relation $\sim$, we define the extended-rank of the pair $(\mathcal{L}(Y), \sim)$ as $\operatorname{acc}(\sim)+\operatorname{rank}\left(\mathcal{L}^{\prime}\left(Y_{I}\right)\right)$ where $\mathcal{L}^{\prime}\left(Y_{I}\right)$ is the reduced form of the equation $\mathcal{L}(Y)$ after applying equivalence relation and the ' $a$ ' restrictions induced by the accidents. Let $\left\{\mathcal{L}_{i}: i \in B\right\}$ be a set of systems of linear equations. Note that for all $i \in B, \mathcal{L}_{i}$ is a system of linear equations. Now we identify the set of systems of linear equations which are actually obtained from different bad cases for three messages $m:=\left(m_{1}, m_{2}, m_{3}\right)$ as shown in Table 5.2. We have another set of single equations indexed by $B^{\prime}$ as shown in Table 6.4. Let $\mathcal{B}_{1}, \mathcal{B}_{2}, \mathcal{B}_{3}, \mathcal{B}_{4}$ denote the set of system of bad equations defined as follows $\mathcal{B}_{1}:=\left\{L_{11}, L_{12}, L_{13}, L_{14}\right\}, \mathcal{B}_{2}:=\left\{L_{21}, L_{22}\right\}, \mathcal{B}_{3}:=\left\{L_{23}, L_{24}\right\}$ (Refer to

$$
\begin{array}{|ll|}
\hline\left(L_{31}\right) & \Sigma_{i}=\Sigma_{j} \\
\left(L_{32}\right) & \Sigma_{i}=X_{j, a} \\
\left(L_{33}\right) & \Theta_{i}=\Theta_{j} \\
\left(L_{34}\right) & \Theta_{i}=X_{j, a} \\
\hline
\end{array}
$$

Table 2. Table representing single bad equations.

Table 5.2), and $\mathcal{B}_{4}:=\left\{L_{31}, L_{32}, L_{33}, L_{34}\right\}$ (Refer to Table 6.4). Let $N_{r, j}$ denote the number of pairs of the form $\left(\sim, \mathcal{L}_{i}\right)$ for some $i \in \mathcal{B}_{j}$, for $j \in\{1,2,3,4\}$ such that $\sim$ is allowed and the extended-rank of the pair is $r$. Then, we have the following general bound for any sum-based construction.
Lemma 9. Let $m=\left(m_{1}, m_{2}, \ldots, m_{q}\right)$ be a $q$-tuple of distinct messages and $t=\left(t_{1}, t_{2}, \ldots, t_{q}\right) \notin \mathcal{V}_{\text {bad }}$. Let $L^{3} \leq 2^{n}$. Then,

$$
\operatorname{Pr}[\Pi \text { is } b a d] \leq O\left(q / 2^{2 n / 3}\right)+q^{3} \epsilon_{e c f}+q^{3} \epsilon_{p c f 1}+q^{4} \epsilon_{p c f 2}+q \epsilon_{e u n i v}
$$

Lemma 10. Let $m=\left(m_{1}, m_{2}, m_{3}\right)$ be a 3-tuple of distinct messages and $t=$ $\left(t_{1}, t_{2}, t_{3}\right) \notin \mathcal{V}_{\text {bad }}$. Let $L^{3} \leq 2^{n}$. Then,
(1) $\epsilon_{e c f} \leq \sum_{r=0}^{4} N_{r, 1} / 2^{n r}$
(2) $\epsilon_{p c f 1} \leq \sum_{r=0}^{4} N_{r, 2} / 2^{n r}$
(3) $\epsilon_{\text {euniv }} \leq \sum_{r=0}^{2} N_{r, 4} / 2^{n(r+1)}$

Moreover, if $m=\left(m_{1}, m_{2}, m_{3}, m_{4}\right)$ be a 4-tuple of distinct messages and $t=$ $\left(t_{1}, t_{2}, t_{3}, t_{4}\right) \notin \mathcal{V}_{\text {bad }}$ then $\epsilon_{p c f 2} \leq \sum_{r=0}^{5} N_{r, 3} / 2^{n r}$.

## $7 \quad$ PRF Security Analysis of 1kf9

In this section we analyze the security of our propopsed construction 1kf9. Mainly we prove the following theorem.

## Theorem 8.

$\mathbf{A d v}_{1 k f 9}^{\operatorname{prf}}(q, \ell, t) \leq \mathbf{A d v}_{E}^{\mathrm{prp}}\left(q, \ell, t^{\prime}\right)+O\left(q l^{2} / 2^{n}+q^{3} l^{4} / 2^{2 n}+q^{4} l^{4} / 2^{3 n}+q^{4} l^{6} / 2^{4 n}\right)$ where $t^{\prime}=t+\mathcal{O}(q L)$ for any L-blocks message $m$.

### 7.1 Revisiting Structure Graph

In this section we revisit the structure graph introduced by Bellare et.al in [4]. We recall that given a $q$-tuple of distinct messages $m$ and a permutation $\pi$, the transcript $\tau(\mathcal{H} \rightarrow \pi)=(x, y)$ represents the set of all inputs and outputs of $\pi$. Here the function $\mathcal{H}$ is nothing but $C B C^{\pi}$. We write $x=\left(x_{i, j}\right)_{(i, j) \in \mathcal{I}}$ and similarly for $y$ where $\mathcal{I}:=\left\{(i, j): i \in[q], j \in\left[L_{i}\right]\right\}$. We have defined an equivalence relation $\sim_{y}$ over $\mathcal{I}$. Let us assume that the permutation $\pi$ does not
map to 0 , i.e., $y_{i, j} \neq 0$ for all $i, j$. Let $\left\{V_{1}, \ldots, V_{s}\right\}$ be the set of all partitions of $\mathcal{I}$ induced by $\sim_{y}$. So $V_{i}$ is a subset of $\mathcal{I}$ whose elements are related to each other by the relation $\sim$. We define a vertex set $V=\left\{V_{0}, V_{1}, \ldots, V_{s}\right\}$. We give an edge from $V_{0}$ to $V_{b}$ if there exists $(i, 1) \in V_{b}$. We also put an edge label $m_{i, 1}$, the first block of the $i^{\text {th }}$ message. Similarly, we give an edge from $V_{a}$ to $V_{b}$ if there exists $(i, j) \in V_{a}$ and $(i, j+1) \in V_{b}$ and we put an edge label $m_{i, j+1}$. We write a labeled edge as $V \xrightarrow{m} V^{\prime}$. It is straightforward to see that the graph is well defined. We call this labeled graph structure graph and denoted $G^{\pi}(m)$. For each message $m_{i}$, we can consider the walk starting from $V_{0}$ to $V_{a}$ for some $a$, following the edge labels $m_{i, 1}, \ldots, m_{i, \ell_{i}}$ one by one. We denote the walk by $W_{i}^{\pi}$ or simply $W_{i}$. Note that the structure graph $G$ would be the union of all walks $W_{i}, 1 \leq i \leq q$.

A node $V$ is said to be a collision node (or true collision) in a structure graph $G$ if the in-degree of the node is at least two. The number of true collision is defined to be the the sum $T C(G):=\sum_{i=1}^{s}\left(\operatorname{indeg}\left(V_{i}\right)-1\right)$.

Definition 10. A collection of edges $C=\left\{V_{i_{1}} \rightarrow V_{i_{2}}, V_{i_{3}} \rightarrow V_{i_{2}}, \ldots, V_{i_{2 k}} \rightarrow\right.$ $\left.V_{i_{1}}\right\}$ in a structure graph $G$ is called an alternating cycle (AC) where $k \geq 2$.

We provide an equivalent definition of the number of accidents of a structure graph as defined in [4].

Definition 11. Let $G_{0}:=G$ be a structure graph. Now we do the following steps until we find an alternating cycle. For $i \geq 1$, we define $G_{i}=G_{i-1} \backslash e$ where $e$ is a labeled edge of an alternating cycle in $G_{i-1}$. Let $G_{t}$ be the final graph (may not be unique as it depends on the choice of the edges from the $A C$ which are removed. The number of accidents of the graph $G_{0}$ is defined to be the number of true collision of $G_{t}$.

One can check that this definition is well defined. In other words, the number of true collision for the final graphs is independent of the choice of the edges removed. We denote the number of accidents and true collision of a structure graph $G^{\pi}(m)$ by $\operatorname{acc}^{\pi}(m)$ and $T C^{\pi}(m)$ respectively.

### 7.2 Characterization of Valid Structure Graphs with 3 and 4 Messages

Definition 12. A Structure Graph $G$ is said to be a Valid Structure Graph, if it meets the following three conditions: (i) $\mid$ Acc $(G) \mid \leq 2$, (ii) No accident within a message $m_{i}$, (iii) At most one accident within three messages $m_{i}, m_{j}, m_{k}$.

### 7.2.1 Important Properties of Valid Structure Graphs for 3 Messages

Lemma 11. A valid structure graph with 3 messages cannot contain an alternating cycle of length 4 .

Proof. Let us consider an alternating cycle $C y c l$ of length at least 4. Let $E_{\text {alt }}:=$ $\{(A B),(A D),(C D),(C B)\}$ be the set of edges of $C y c l$ as shown in Fig. 7.1. Now we make the following two important observations :


Fig. 7.1. Alternating cycle of length 4
(i) As we have three messages, at least one message covers two edges from $E_{\text {alt }}$.
Without loss of generality let $m_{i}$ be the message that covers two edges.
(ii) The covered edges will be parallel, otherwise there will be an accident within the walk of $m_{i}$.
W.l.o.g, let the covered edges of $m_{i}$ be $(A B)$ and ( $\left.C D\right)$. Let $m_{j}$ covers consider the message which covers the edge $(C B)$. W.l.o.g, let it be $m_{j}$. Now to cover that edge, $m_{j}$ could come to node $C$ in either of the two ways :
(a) $m_{j}$ follows the walk of $m_{i}$ and reaches to $C$
(b) $m_{j}$ does not follow the walk of $m_{i}$.

For case (a) when $m_{j}$ covers the edge $(C B)$, then there will be an accident within the walk of $m_{j}$. For case (b) when $m_{j}$ covers the edge $(C B)$ then $m_{i}, m_{j}$ will collide twice and hence the number of accident in $\left(m_{i}, m_{j}\right)$ pair will be 2 . As, in both the cases the condition for a valid structure graph is violated, the result follows.

Lemma 12. A valid structure graph with 3 messages cannot contain an alternating cycle of length 6 .

Proof. Let $\mathrm{Cycl}_{6}$ be an alternating cycle of length 6 in the valid structure graph $G$ with 3 messages. Let $m_{1}$ be the message taking part in two collision points say $C_{1}$ and $C_{2}$. Now consider other messages (say $m_{2}$ and $m_{3}$ ) taking part in these collisions, i.e. $C_{1}=\operatorname{coll}\left(m_{1}, m_{2}\right), C_{2}=\operatorname{coll}\left(m_{1}, m_{3}\right)$. Now it is easy to see that there are 2 accidents in $m_{1}, m_{2}$ and $m_{3}$ that violates the validity of a structure graph. Hence no valid graph is possible with 6 -alternating cycle.

### 7.2.2 Important Properties of Valid Structure Graphs for 4 Messages

Claim 1 For any 4-length alternating cycle in a valid structure graph with 4 messages, the 4 edges must come from distinct messages

Proof. If not, then 3 distinct messages cover 4 edges of the 4-length alternating cycle. But according to Lemma 11, a valid structure graph with 3 messages cannot contain a 4 -length alternating cycle.

Lemma 13. A valid structure graph with 4 messages cannot contain a 4-length alternating cycle with number of accidents 2 .

Proof. Due to Claim 1 without loss of generality, we can assume that the edges $\mathrm{AB}, \mathrm{AD}, \mathrm{CB}$ and CD of an alternating cycle belong to messages $m_{1}, m_{2}, m_{3}, m_{4}$ respectively, where $m_{1}$ and $m_{3}$ have an accident at B and $m_{2}$ and $m_{4}$ meet at B to close the alt-cycle with an induced collision. Now, if there is a second accident, it cannot involve any one of $m_{1}$ or $m_{3}$, otherwise it will violate condition 2 (\#acc at most one with any 3 messages). Thus, the second accident, if any, must involve $m_{2}$ and $m_{4}$. But again this is not allowed, since $m_{2}$ and $m_{4}$ has already collided at B.

Lemma 14. A valid structure graph with 4 messages cannot contain multiple alternating cycle of length 4 .

Proof. Due to Claim 1, without loss of generality, we can assume that the edges $\mathrm{AB}, \mathrm{AD}, \mathrm{CB}$ and CD of an alternating cycle belong to messages $m_{1}, m_{2}, m_{3}, m_{4}$ respectively. Now, if another 4 -alternating cycle exists, Claim 1 must hold for this second cycle as well. This implies that two edges (from two different messages) must be shared between the two cycles. The shared edges may be any one of the 4 pairs from $\mathrm{AB}, \mathrm{AD}, \mathrm{CB}$ and CD . Case a) Pairs that do not have a common node from $\mathrm{A}, \mathrm{B}, \mathrm{C}, \mathrm{D}$, i.e., pair $(\mathrm{AB}, \mathrm{CD})$ or pair $(\mathrm{AD}, \mathrm{BC})$ : Then the other two edges of the second cycle will add two more accidents, one in node B and another in node C, violating condition 3. Case b) Pairs that have a common node. In this case, two possible graphs are possible, as shown in the diagram. The other two edges must meet at a fifth node, say E. [Show the table].

Lemma 15. A valid structure graph with 4 messages cannot contain an alternating cycle of length 6 .

Proof. Let $\mathrm{Cycl}_{6}$ be the alternating cycle of length 6 in the valid structure graph $G$ with 4 messages. As there are 3 accident points $C_{1}, C_{2}, C_{3}$ in $C y c l_{6}$, there will be at least one message say $m_{1}$ taking part in two collision points say $C_{1}$ and $C_{2}$. Now consider other messages (say $m_{2}$ and $m_{3}$ ) taking part in these collisions, i.e. $C_{1}=\operatorname{coll}\left(m_{1}, m_{2}\right), C_{2}=\operatorname{coll}\left(m_{1}, m_{3}\right)$. Now it is easy to see that there are 2 accidents in $m_{1}, m_{2}$ and $m_{3}$ that violates the validity of a structure graph. Hence no valid graph is possible with 6-length alternating cycle
7.2.3 List of Valid Structure Graphs with $\mathbf{3}$ and 4 messages Given all the properties, now we list down all the possible structure graphs with 3 and 4 messages as follows:
(I) Acc $=\mathbf{0}$ for 3 messages: As no accident is present, the only possible structure graph has the following structure depicted in Fig. 7.2:
(II) $\mathbf{A c c}=\mathbf{1}$ for 3 messages: From Lemma 11, we observe that, there can


Fig. 7.2. Structure graph of 3 messages with $A c c=0$
be no valid graph 4-length alternating cycles. So we consider structure graphs where number of true-collision is 1 and the graph is shown in Fig 7.3.
$(\mathbf{I I I})$ Acc $=\mathbf{0}$ for 4 messages As no accident is present, the only possible


Fig. 7.3. Structure graph of 3 messages with $A c c=1$ (at node $C$ )
structure graph has the following structure depicted in Fig. 7.4:
(IV) Acc $=\mathbf{1}$ for 4 messages: We can have two types of graph in this case:


Fig. 7.4. Structure graph of 4 messages with $A c c=0$

- 1 accident with 1 collision point: This graph is shown in Fig. 7.5.
- Graph with 1-alternating cycle: This graph is shown in Fig. 7.6 .


Fig. 7.5. Structure graph of 4 messages with $A c c=1$ (at node $C$ )
$(V)$ Acc $=\mathbf{2}$ for 4 messages: From Lemma 13 and 15, we observe that, there can be no valid graphs with alternating 4 -cycle or alternating 6 cycle. Hence there is only one possible structure graph - with one accident $C_{1}$ occuring between two messages (say $m_{1}$ and $m_{2}$ ) and the other accident $C_{2}$ occuring for the remaining messages (here $m_{3}$ and $m_{4}$ ). This graph also satisfy the condition: $\left(m_{1}, m_{2}\right)$ and ( $m_{3}, m_{4}$ ) doesn't meet after collision $C_{1}$ and $C_{2}$ respectively as depicted in Fig. 7.7.


Fig. 7.6. Structure graph of 4 messages with $A c c=1$ (at node $B$ ) and an induced collision (at node $D$ )


Fig. 7.7. Structure graph of 4 messages with $A c c=2$ (at nodes $E$ and $F$ )

## 8 Rank Analysis of Systems of Equations for Bad Cases

### 8.1 Rank Analysis of Fully Covered Bad Equations

### 8.1.1 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=\Sigma_{j}, \Theta_{i}=\Theta_{k}\right)$ for $A c c=$

 0 and $A c c=1$.Case (a) When $A c c=0$, then $\Sigma_{i}=\Sigma_{j}$ implies $\alpha Y_{i, l_{i}}+\alpha Y_{j, l_{j}}=O^{n-1} 1$. Let us aussme that $p$ is the length of the longest common prefix of $M_{i}$ and $M_{k}$ and wihput loss of generality $l_{i}>l_{k}$. Therefore, we have following equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+\alpha Y_{j, l_{j}}=0^{n-1} 1  \tag{8}\\
& Y_{i, p+1}+\ldots Y_{i, l_{i}}+Y_{k, p+1}+\ldots Y_{k, l_{j}}=0 \tag{9}
\end{align*}
$$

Now it is to be noted that, if $M_{k}$ is a prefix of $M_{i}$, then $Y_{i, p+1}+\ldots Y_{i, l_{i}}$ contains at least 3 variables. Therefore, $Y_{j, l_{j}}$ could be equal to one of these three variables, and other two variables will remain free. In that case we will identify one such variable $Y_{i, s}$ which is not equal to $Y_{j, l_{j}}$ and choose $Y_{i, l_{i}}$. If $M_{k}$ is not a prefix of $M_{i}$ then $Y_{i, p+1}+\ldots Y_{i, l_{i}}+Y_{k, p+1}+\ldots Y_{k, l_{j}}$ conatins at least 3 variables and therefore, $Y_{j, l_{j}}$ could be equal to one of these three variables; we will identify one of the remaining free variable $Y_{i, s}$ which is not equal to $Y_{j, l_{j}}$ and choose $Y_{i, l_{i}}$. Therefore we identify two such variables, one in each equation, giving us rank 2. Case (b) When $A c c=1$, we argue that rank of $\mathcal{L}(Y)$ will be 2 . For $A c c=1$, we introduce one more equation

$$
\begin{equation*}
Y_{i, \beta}+Y_{j, \gamma}=m \tag{10}
\end{equation*}
$$

along with Equation (8) and (9). Note that if $A c c=1$, then $\Sigma_{i}=\Sigma_{j}$ implies eiher of the following two cases: (i) $\alpha Y_{i, l_{i}}=\alpha Y_{j, l_{j}}$. or (ii) $\alpha Y_{i, l_{i}} \neq \alpha Y_{j, l_{j}}$ but fix0 $\left(\alpha Y_{i, l_{i}}\right)=\operatorname{fix} 0\left(\alpha Y_{j, l_{j}}\right)$. Note that, considering case (i), this is equivalent to considering the equation $Y_{i, \beta}+Y_{j, \gamma}=m$. According to our assumption $p$ be the last index where $M_{i}$ and $M_{k}$ is identical. Therefore, as argued before, $Y_{i, p+1}+\ldots Y_{i, l_{i}}+Y_{k, p+1}+\ldots Y_{k, l_{j}}$ contains at least three variables. Now $Y_{j, \gamma}$ could be equal to any one of the three variables; thus we will be left with at least two variable which are free . Let us consider $Y_{i, s} \neq Y_{j, \gamma}$. Therefore we identify two free variables $Y_{i, \beta}$ and $Y_{i, s}$, one in each equation, giving us rank 2. If case (ii) occurs then we consider the Equation (8). In that case $Y_{j, \gamma}$ and $Y_{j, l_{j}}$ could be equal to any two of the three variables. Then also we will be left with at least one variable $Y_{i, s}$. Therefore, we identify two free variables $Y_{i, \beta}$ and $Y_{i, s}$, one in each equation, such that the rank becomes 2 .
8.1.2 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=X_{j, r}, \Theta_{i}=\Theta_{k}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): When \#Acc $=0$, then we argue that rank of $\mathcal{L}(Y)$ is 2 . We have the following two equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+Y_{j, r-1}+M^{j, r}=0  \tag{11}\\
& Y_{i, p+1}+\ldots Y_{i, l_{i}}+Y_{k, p+1}+\ldots Y_{k, l_{k}}=0 \tag{12}
\end{align*}
$$

where $p$ is the length of the longest common prefix of $M_{i}$ and $M_{k}$. It is to be noted that there are at least three distinct variables in Equation (12). Now, we identify $Y_{i, l_{i}}$ and one of the remaining free variable $Y_{i, s}$ out of above three variables which is distinct from $Y_{i, l_{i}}$ and $Y_{j, r-1}$, giving us rank 2. Case (b): When $A c c=1$, then one additional equation

$$
\begin{equation*}
Y_{i, \beta}+Y_{j, \gamma}=m \tag{13}
\end{equation*}
$$

is introduced. Now if $Y_{i, \beta} \neq Y_{i, l_{i}}$ and $Y_{j, \gamma} \neq Y_{i, l_{i}}$, then we identify two variables $Y_{i, \beta}$ and $Y_{i, l_{i}}$ such that rank of $\mathcal{L}(Y)$ with $A c c=1$ is 2 . If this is not the case, we identify $Y_{i, \beta}$ and $Y_{i, s}$ which is one of the out of three variables in Equation (12), such that the rank becoms 2 again.
8.1.3 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=\Sigma_{j}, \Theta_{i}=X_{k, r}\right)$ for $A c c=0$ and $A c c=1$
Case (a): Let us first consider that $A c c=0$. Now we have the following two equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+\alpha Y_{j, l_{j}}=0^{n-1} 1  \tag{14}\\
& \alpha\left(Y_{i, 0}+Y_{i, 1}+\ldots Y_{i, l_{i}}\right)=Y_{k, r-1}+m_{r} \tag{15}
\end{align*}
$$

From Equation (14) and (15), we identify two free variables $Y_{i, l_{i}}$ and $Y_{i, 0}$, giving us rank 2. Case(b): When $A c c=1$, then along with Equation (14) and (15), we have an additional equation

$$
Y_{i, \beta}+Y_{j, \gamma}=m
$$

Now, $\Sigma_{i}=\Sigma_{j}$ can occur in either of the following ways: (i) $\alpha Y_{i, l_{i}}=\alpha Y_{j, l_{j}}$ or (ii) $\alpha Y_{i, l_{i}} \neq \alpha Y_{j, l_{j}}$ but fix0 $\left(\alpha Y_{i, l_{i}}\right)=$ fix0 $\left(\alpha Y_{j, l_{j}}\right)$. Note that, considering case (i) is equivalent to considering the equation $Y_{i, \beta}+Y_{j, \gamma}=m$. Therefore we identify two free variables $Y_{i, 0}$ and $Y_{i, \beta}$, such that the rank becomes 2. Considering case (ii) is boiling down to considering Equation (14). Therefore, we identify $Y_{i, l_{i}}$ and $Y_{i, 0}$, such that the rank becomes 2 again.
8.1.4 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=X_{j, s}, \Theta_{i}=X_{k, r}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): Let us consider $A c c=0$. We have the following equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+Y_{j, s-1}=m^{*}  \tag{16}\\
& \alpha\left(Y_{i, 0}+Y_{i, 1}+\ldots Y_{i, l_{i}}\right)=Y_{k, r-1}+m^{* *} \tag{17}
\end{align*}
$$

In this case we identify two free variables $Y_{i, 0}$ and $Y_{i, l_{i}}$. Case (b): When $A c c=1$, we have an additional equation

$$
Y_{i, \beta}+Y_{j, \gamma}=m
$$

Thus, again we can identify two free variables $Y_{i, 0}$ and $Y_{i, l_{i}}$ and the rank does not decrease.

### 8.2 Rank Analysis of Single Equations

8.2.1 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=\Sigma_{j}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): For Acc $=0, \Sigma_{i}=\Sigma_{j}$ implies $\alpha Y_{i, l_{i}}+\alpha Y_{j, l_{j}}=0^{n-1} 1$. Since $Y_{i, l_{i}}$ is not trivially equal to $Y_{j, l_{j}}, \mathcal{L}(Y)$ will have rank 1 for choosing variable $Y_{i, l_{i}}$.
Case (b): For $A c c=1, \Sigma_{i}=\Sigma_{j}$ implies either (i) $\alpha Y_{i, l_{i}}+\alpha Y_{j, l_{j}}=0^{n-1} 1$ or (ii) $\alpha Y_{i, l_{i}}=\alpha Y_{j, l_{j}}$ but fix0 $\left(\alpha Y_{i, l_{i}}\right)=$ fixo $\left(\alpha Y_{j, l_{j}}\right)$. Therefore, considering case (ii) boils down to considering the Equation (18) which is induced by the accident.

$$
\begin{equation*}
Y_{i, \beta}+Y_{j, \gamma}=m \tag{18}
\end{equation*}
$$

Therefore, choosing $Y_{i, \beta}$ gives the rank of $\mathcal{L}(Y)$ to be 1 .
8.2.2 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=X_{j, r}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): For $A c c=0$, we choose $Y_{i, l_{i}}$ such that rank of $\mathcal{L}(Y)$ is 1 as equality of $\Sigma_{i}$ and $X_{j, r}$ is not trivial equality.
Case (b): For $A c c=1$, we introduce the collision relation $Y_{i, \beta}+Y_{j, \gamma}=m$. Since any accident gives a linearly indpendent equation, therefore we choose $Y_{i, \beta}$ to show the rank of $\mathcal{L}(Y)$ with $A c c=1$ is 1 .
8.2.3 Calculating the rank of $\mathcal{L}(Y)=\left(\Theta_{i}=X_{k, r}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): For $A c c=0$, we choose $Y_{i, 0}$ such that rank of $\mathcal{L}(Y)$ is 1 as equality of $\Theta_{i}$ and $X_{k, r}$ is not trivial equality.
Case (b): For $A c c=1$, we introduce the collision relation $Y_{i, \beta}+Y_{k, \gamma}=m$. Since any accident gives a linearly indpendent equation, therefore we choose $Y_{i, \beta}$ to show the rank of $\mathcal{L}(Y)$ with $A c c=1$ is 1 .
8.2.4 Calculating the rank of $\mathcal{L}(Y)=\left(\Theta_{i}=\Theta_{k}\right)$ for Acc $=0$ and Acc $=1$.
Case (a): Let $p$ be the longest common prefix of $M_{i}$ and $M_{j}$. Therefore, $\Theta_{i}=\Theta_{k}$ gives the following equation

$$
\begin{equation*}
Y_{i, p+1}+\ldots Y_{i, l_{i}}+Y_{j, p+1}+\ldots Y_{j, l_{j}}=0 \tag{19}
\end{equation*}
$$

Note that there must be at least three distinct variables in Equation (19). Therefore, for $A c c=0$, we choose any of the three variables $Y_{i, s}$ such that rank of $\mathcal{L}(Y)$ is 1 .
Case (b): For $A c c=1$, we introduce the collision relation $Y_{i, \beta}+Y_{k, \gamma}=m$. Since any accident gives a linearly indpendent equation, therefore we choose $Y_{i, \beta}$ to show the rank of $\mathcal{L}(Y)$ with $A c c=1$ is 1 .

### 8.3 Rank Analysis of Pseudo-Covered Bad Equations with 3 Messages

8.3.1 Calculating the rank of $\mathcal{L}(Y)=\left(\Sigma_{i}=X_{j, r}, Y_{j, r}+Y_{*}=t_{i}\right)$ for $A c c=0$ and $A c c=1$.
Case (a): Let us consider $A c c=0$. We have the following Equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+Y_{j, r-1}=m_{j, r}  \tag{20}\\
& Y_{j, r}+Y_{*}=t_{i} \tag{21}
\end{align*}
$$

We identify two variables $Y_{j, r}$ and $Y_{j, r-1}$ such that the contribution matrix $E$ becomes non-singular. It is easy to note that $Y_{j, r}$ can never be equal to $Y_{j, r-1}$ as we are not allowing any loop in the structure graph.
Case (b): When $A c c=1$, we additionally introduce one more equation

$$
Y_{i, s}+Y_{j, t}=m
$$

We identify the same two variables $Y_{j, r}$ and $Y_{j, r-1}$ such that one can show the rank of $\mathcal{L}(Y)$ with $A c c=1$ is 2 .
8.3.2 Calculating the rank of $\mathcal{L}(Y)=\left(\Theta_{i}=X_{j, r}, Y_{j, r}+Y_{*}=t_{i}\right)$ for $A c c=0$ and $A c c=1$.
One can argue the rank of $\mathcal{L}(Y)$ for $A c c=0$ and $A c c=1$ is 2 in the same line of argument for the rank analysis of the previous case.

### 8.4 Rank Analysis of Pseudo-Covered Bad Equations with 4 Messages

8.4.1 Calculating the rank of $\mathcal{L}(\boldsymbol{Y})=\left(\Sigma_{i}=X_{k, e}, \Sigma_{j}=X_{l, f}, Y_{k, e}+\right.$ $Y_{l, f}=t_{i}+t_{j}$ for $A c c=0,1$ and 2$)$.
Case (a): Let us consider $A c c=0$. We have the following equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+Y_{k, e-1}=m^{*}  \tag{22}\\
& \alpha Y_{j, l_{j}}+Y_{l, f-1}=m^{* *}  \tag{23}\\
& Y_{k, e}+Y_{l, f}=t_{i}+t_{j} \tag{24}
\end{align*}
$$

Now we analyse the rank in three cases. Case (i) when $Y_{k, e} \neq Y_{i, l_{i}}$ and $Y_{k, e} \neq Y_{j, l_{j}}$ then we identify three variables $Y_{i, l_{i}}, Y_{j, l_{j}}$ and $Y_{k, e}$ such that the rank of $\mathcal{L}(Y)$ is 3 .

Case (ii) when $Y_{l, f} \neq Y_{i, l_{i}}$ and $Y_{l, f} \neq Y_{j, l_{j}}$ then we can identify the variables $Y_{i, l_{i}}, Y_{j, l_{j}}$ and $Y_{l, f}$ such that the rank will become 3.

Case (iii) If none of the above two cases occur (i.e., $Y_{i, l_{i}}=Y_{k, e}, Y_{j, l_{j}}=Y_{l, f}$ ) then we identify three variables $Y_{i, l_{i}}, Y_{j, l_{j}}$ and $Y_{k, e-1}$ such that the rank becomes 3.

Case (b): When $A c c=1$ we introduce Equation (35) along with the previous three equations.

$$
\begin{equation*}
Y_{i, s}+Y_{j, t}=m \tag{25}
\end{equation*}
$$

Even if $Y_{i, s}$ or $Y_{j, t}$ is equal to any of the previously chosen free variables, the rank does not decrease.
Case (c): When $A c c=2$, we introduce an additional equation, namely, Equation (36) as below.

$$
\begin{equation*}
Y_{k, s^{\prime}}+Y_{l, t^{\prime}}=m^{\prime} \tag{26}
\end{equation*}
$$

According to our assumptions, the second accident must occur between two other messages that were not involved in the first accident. Hence, we can choose an additional free variable and hence the rank becomes 4 .
8.4.2 Calculating the rank of $\mathcal{L}(\boldsymbol{Y})=\left(\Sigma_{i}=X_{k, e}, \Theta_{j}=X_{l, f}, Y_{k, e}+\right.$ $Y_{l, f}=t_{i}+t_{j}$ for $A c c=0,1$ and 2$)$.
Case (a): Let us consider $A c c=0$. We have the following Equations:

$$
\begin{align*}
& \alpha Y_{i, l_{i}}+Y_{k, e-1}=m^{*}  \tag{27}\\
& \alpha\left(Y_{j, 0}+Y_{j, 1}+\ldots+Y_{j, l_{j}}\right)+Y_{l, f-1}=m^{* *}  \tag{28}\\
& Y_{k, e}+Y_{l, f}=t_{i}+t_{j} \tag{29}
\end{align*}
$$

Now we analyse the rank in three cases. Case (i) when $Y_{k, e} \neq Y_{i, l_{i}}$ and $Y_{k, e} \neq Y_{j, 0}$ then we identify three variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{k, e}$ such that the rank of $\mathcal{L}(Y)$ is 3.

Case (ii) when $Y_{l, f} \neq Y_{i, l_{i}}$ and $Y_{l, f} \neq Y_{j, 0}$ then we can identify the variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{l, f}$ such that the rank will become 3.

Case (iii) If none of the above two cases occur (i.e., $Y_{i, l_{i}}=Y_{k, e}, Y_{j, 0}=Y_{l, f}$ ) then we identify three variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{k, e-1}$ such that the rank becomes 3.

Case (b): When $A c c=1$ we introduce Equation (35) along with the previous three equations.

$$
\begin{equation*}
Y_{i, s}+Y_{j, t}=m \tag{30}
\end{equation*}
$$

Even if $Y_{i, s}$ or $Y_{j, t}$ is equal to any of the previously chosen free variables, the rank does not decrease.
Case (c): When $A c c=2$, we introduce an additional equation, namely, Equation (36) as below.

$$
\begin{equation*}
Y_{k, s^{\prime}}+Y_{l, t^{\prime}}=m^{\prime} \tag{31}
\end{equation*}
$$

According to our assumptions, the second accident must occur between two other messages that were not involved in the first accident. Hence, we can choose an additional free variable and hence the rank becomes 4.
8.4.3 Calculating the rank of $\mathcal{L}(\boldsymbol{Y})=\left(\Theta_{i}=X_{k, e}, \Sigma_{j}=X_{l, f}, Y_{k, e}+\right.$ $Y_{l, f}=t_{i}+t_{j}$ for $A c c=0,1$ and 2$)$.
This case is similar to the previous case, where $\mathcal{L}(Y)=\left(\Sigma_{i}=X_{k, e}, \Theta_{j}=\right.$ $X_{l, f}, Y_{k, e}+Y_{l, f}=t_{i}+t_{j}$.
8.4.4 Calculating the rank of $\mathcal{L}(Y)=\left(\Theta_{i}=X_{k, e}, \Theta_{j}=X_{l, f}, Y_{k, e}+\right.$ $Y_{l, f}=t_{i}+t_{j}$ for $A c c=0,1$ and 2$)$.
Case (a): Let us consider $A c c=0$. We have the following Equations:

$$
\begin{align*}
& \alpha\left(Y_{i, 0}+Y_{i, 1}+\ldots+Y_{i, l_{i}}\right)+Y_{k, e-1}=m^{*}  \tag{32}\\
& \alpha\left(Y_{j, 0}+Y_{j, 1}+\ldots+Y_{j, l_{j}}\right)+Y_{l, f-1}=m^{* *}  \tag{33}\\
& Y_{k, e}+Y_{l, f}=t_{i}+t_{j} \tag{34}
\end{align*}
$$

Now we analyse the rank in three cases. Case (i) when $Y_{k, e} \neq Y_{i, l_{i}}$ and $Y_{k, e} \neq Y_{j, 0}$ then we identify three variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{k, e}$ such that the rank of $\mathcal{L}(Y)$ is 3.

Case (ii) when $Y_{l, f} \neq Y_{i, l_{i}}$ and $Y_{l, f} \neq Y_{j, 0}$ then we can identify the variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{l, f}$ such that the rank will become 3.

Case (iii) If none of the above two cases occur (i.e., $Y_{i, l_{i}}=Y_{k, e}, Y_{j, 0}=Y_{l, f}$ ) then we identify three variables $Y_{i, l_{i}}, Y_{j, 0}$ and $Y_{k, e-1}$ such that the rank becomes 3.

Case (b): When $A c c=1$ we introduce Equation (35) along with the previous three equations.

$$
\begin{equation*}
Y_{i, s}+Y_{j, t}=m . \tag{35}
\end{equation*}
$$

Even if $Y_{i, s}$ or $Y_{j, t}$ is equal to any of the previously chosen free variables, the rank does not decrease.
Case (c): When $A c c=2$, we introduce an additional equation, namely, Equation (36) as below.

$$
\begin{equation*}
Y_{k, s^{\prime}}+Y_{l, t^{\prime}}=m^{\prime} \tag{36}
\end{equation*}
$$

According to assumptions, the second accident must occur between two other messages that were not involved in the first accident. Hence, we can choose an additional free variable and hence the rank becomes 4.

## 9 PRF Security Analysis of 1kPMAC_Plus

### 9.1 Preparation

Taking advantage of Theorem 7, to prove the PRF security of 1kPMAC_Plus, we need to upper bound its three items, extended-cover-free $\epsilon_{e c f}$, pseudo-cover-free $\epsilon_{p c f}$, and block-wise universal $\epsilon_{a u}$.

To show they are sufficiently small, we would define some bad events on inputs to block ciphers. Each bad event is equivalent to a equation set over block cipher outputs as variables. By solving the equations we get an upper bound of permutations over $\{0,1\}^{n}$ that can induce the bad events. Then notice that there are totally $2^{n}$ ! permutations, we get the occurrence probability for each bad event.

1. $\exists X_{i, l} \in\left\{\mathrm{Cst}_{1}, \mathrm{Cst}_{2}\right\}$, for some $i \in[q]$ and $l \in\left[\ell_{i}\right]$. This implies no more than $\sum_{j=1}^{2} \sum_{i=1}^{q} \sum_{l=1}^{\ell_{i}}$ equations of the form,

$$
X_{i, l}=M_{i, l} \oplus 2^{l-1} \Delta_{1} \oplus 2^{2 l-2} \Delta_{2}=\text { Cst }_{j}
$$

Notice that $\Delta_{1}=\pi\left(\mathrm{Cst}_{1}\right)$ and $\Delta_{2}=\pi\left(\mathrm{Cst}_{2}\right)$, we have no more than $\left(2^{n}-1\right)$ ! permutations satisfying each equation, and totally we have at most $\varsigma_{-1}=\sum_{j=1}^{2} \sum_{i=1}^{q} \sum_{l=1}^{\ell_{i}}\left(\left(2^{n}-1\right)!\right)$ permutations over $\{0,1\}^{n}$. Then, the nonoccurrence of this event ensures the $\Delta_{1}, \Delta_{2}$ values are independent of $Y_{i, l}$ values.
2. $\exists X_{i 1, l 1}=X_{i 2, l 2}=X_{i 3, l 3}$ for some $i 1, i 2, i 3 \in[q]$ and distinct $l 1, l 2, l 3 \in[\ell]$. This implies no more than $\binom{q \ell}{3}$ equations of the form,

$$
\left[\begin{array}{l}
2^{l 1-1} \oplus 2^{l 2-1}, 2^{2(l 1-1)} \oplus 2^{2(l 2-1)} \\
2^{l 1-1} \oplus 2^{l 3-1}, 2^{2(l 1-1)} \oplus 2^{2(l 3-1)}
\end{array}\right] \times\left[\begin{array}{l}
\Delta_{1} \\
\Delta_{2}
\end{array}\right]=\left[\begin{array}{l}
M_{i 1, l 1} \oplus M_{i 2, l 2} \\
M_{i 1, l 1} \oplus M_{i 3, l 3}
\end{array}\right]
$$

The determinant of its coefficient matrix is $\left(2^{l 1-1} \oplus 2^{l 2-1}\right)\left(2^{l 1-1} \oplus 2^{l 3-1}\right)$ $\left(2^{l 2-1} \oplus 2^{l 3-1}\right) \neq 0^{n}$ for any distinct $l 1, l 2, l 3$, so this matrix has rank $=2$ and we have $\left(2^{n}-2\right)$ ! solutions on $\Delta_{1}$ and $\Delta_{2}$ for each equation. Then by this we get at most $\varsigma_{0}=\binom{q \ell}{3}\left(\left(2^{n}-2\right)!\right) \leq q^{3} \ell^{3} / 6\left(\left(2^{n}-2\right)!\right)$ permutations from Perm $(n)$.

Based on the above, let us formally upper bound the three items. In each case, we show how to find a rank $=2$ coefficients matrix.

### 9.2 Upper Bounding extended-cover-free $\epsilon_{\text {ecf }}$

According to the definition of extended-cover-freeness, we have 9 bad events in this case, because in the previous inputs to block ciphers, we have both $\Delta_{1}=$ $\pi\left(\mathrm{Cst}_{1}\right), \Delta_{2}=\pi\left(\mathrm{Cst}_{2}\right)$, and $Y_{i, l}=\pi\left(X_{i, l}\right)$, as listed in Table. 4.

1. $\exists \Sigma_{i}=\operatorname{Cst}_{j 1}$ for some $j 1 \in[2]$ and $\Theta_{i}=\operatorname{Cst}_{j 2}$ for some $j 2 \in[2]$. This implies

$$
\left[\begin{array}{cc}
1, & 1, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \\
\cdots, & \cdots, 2^{1}
\end{array}\right] \times\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}\right]^{T}=\left[\begin{array}{c}
\operatorname{Cst}_{j 1} \\
\operatorname{Cst}_{j 2}
\end{array}\right]
$$

Let us analyze in more detail.
(a) If $\ell_{i}=1$ and $\mathrm{Cst}_{j 2}=2$ Cst $_{j 1}$. We get only one equation $Y_{i, 1}=\operatorname{Cst}_{j 1}$, and for $q$ messages, we have at most $\sum_{i=1}^{q}\left(\left(2^{n}-1\right)!\right)$ permutations.
(b) Else if $\ell_{i}=1$ and $\operatorname{Cst}_{j 2} \neq 2 \mathrm{Cst}_{j 1}$. There is no solution.
(c) Else if $\ell_{i} \geq 2$, and $\nexists Y_{i, l 1}=Y_{i, l 2}$ for any distinct $l 1, l 2 \in\left[\ell_{i}\right]$. Then we have a non-singular submatrix $\left[1,1 ; 2^{2}, 2^{1}\right]$ on the left side. For any values of $Y_{i, l}(l \geq 3)$, we have a unique solution for $Y_{1}^{i}$ and $Y_{2}^{i}$. For $q$ messages, we have at most $\sum_{i=1}^{q} \sum_{j 1=1}^{2} \sum_{j 2=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations in total.
(d) Else $\ell_{i} \geq 2$, and $\exists Y_{i, l 1}=Y_{i, l 2}$ for distinct $l 1, l 2 \in\left[\ell_{i}\right]$. We have an equation $\left(2^{l 1-1} \oplus 2^{l 2-1}\right) \Delta_{1} \oplus\left(2^{2(l 1-1)} \oplus 2^{2(l 2-1)}\right) \Delta_{2}=M_{l 1}^{i} \oplus M_{l 2}^{i}$, and an equation set of the form

$$
\left[\begin{array}{cc}
1 \oplus 1, & \cdots \\
2^{\ell_{i}-l 1+1} \oplus 2^{\ell_{i}-l 2+1}, & \cdots
\end{array}\right] \times\left[Y_{i, l 1}, \cdots\right]^{T}=\left[\begin{array}{c}
\operatorname{Cst}_{j 1} \\
\operatorname{Cst}_{j 2}
\end{array}\right] .
$$

Since $2^{\ell_{i}-l 1+1} \oplus 2^{\ell_{i}-l 2+1} \neq 0^{n}$, for any values of $Y_{i, l}(l \neq l 1, l 2)$, we have at most one value for $Y_{i, l 1}$. By the independence of $\Delta_{1}, \Delta_{2}$ and $Y_{i, l}$. In total we have at most $\sum_{i=1}^{q}\binom{\ell_{i}}{2} \sum_{j 2=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations.

To summarize Case 1, we have at most $\varsigma_{1}=\left(q \ell^{2}+q\left(2^{n}-1\right)\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
2. $\exists \Sigma_{i}=$ Cst $_{j}$ for some $j \in[2]$ and $\Theta_{i}=X_{u, v}$ for some $u \in[q], v \in\left[\ell_{u}\right]$. This implies

$$
\left[\begin{array}{ccc}
1, & 1, & \cdots, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, & \cdots, \\
2^{1}, & 2^{v-1}, & 2^{2(v-1)}
\end{array}\right] \times\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, \Delta_{1}, \Delta_{2}\right]^{T}=\left[\begin{array}{c}
\mathrm{Cst}_{j} \\
M_{u, v}
\end{array}\right] .
$$

By the independence of $\Delta_{1}, \Delta_{2}$ and $Y_{i, l}$, let us analyze in detail.
(a) If $\nexists Y_{i, l 1}=Y_{i, l 2}$ for any distinct $l 1, l 2 \in\left[\ell_{i}\right]$. The coefficient matrix on the left side has a non-singular submatrix $\left[1,0 ; 2^{1}, 2^{v-1}\right]$. For $q$ messages, we have at most $\sum_{i=1}^{q} \sum_{l=1}^{\ell_{i}} \sum_{u=1}^{q} \sum_{v=1}^{\ell_{u}} \sum_{j=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations.
(b) Else $\exists Y_{i, l 1}=Y_{i, l 2}$ for distinct $l 1, l 2 \in\left[\ell_{i}\right]$. Then, we have one equation over $\Delta_{1}$ and $\Delta_{2}$ by the 2-collision, and another equation over $Y_{i, l 1}$ (with coefficient $\left.2^{\ell_{i}-l 1+1} \oplus 2^{\ell_{i}-l 2+1} \neq 0\right), \Delta_{1}$ and $\Delta_{2}$. By their independence, we have at most $\sum_{i=1}^{q}\binom{\ell_{i}}{2} \sum_{u=1}^{q} \sum_{v=1}^{\ell_{u}}\left(\left(2^{n}-2\right)!\right)$ permutations.
To summarize Case 2, we have at most $\varsigma_{2}=\left(2 q^{2} \ell^{2}+q^{2} \ell^{3} / 2\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
3. $\exists \Sigma_{i}=X_{u, v}$ for some $u \in[q], v \in\left[\ell_{u}\right]$ and $\Theta_{i}=\operatorname{Cst}_{j}$ for some $j \in[2]$. This implies

$$
\left.\left[\begin{array}{ccc}
1, & 1, & \cdots, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \cdots, 2^{v-1}, & 0
\end{array}\right] \quad 2^{2(v-1)}\right] \times\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, \Delta_{1}, \Delta_{2}\right]^{T}=\left[\begin{array}{c}
M_{u, v} \\
\mathrm{Cst}_{j}
\end{array}\right]
$$

The analysis is similar with Case 2 , and we have at most $\varsigma_{3}=\left(2 q^{2} \ell^{2}+\right.$ $\left.q^{3} \ell^{3}\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
4. $\exists \Sigma_{i}=X_{u 1, v 1}$ and $\Theta_{i}=X_{u 2, v 2}$ for some $i, u 1, u 2 \in[q], v 1 \in\left[\ell_{u 1}\right], v 2 \in\left[\ell_{u 2}\right]$. Then we have an equation set

$$
\left[\begin{array}{cc}
1, & 1, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \\
, & \cdots, \\
2^{1}, & 2^{v 2-1}, \\
2^{v 1-1} & 2^{2(v 2-1)}
\end{array}\right] \times\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, \Delta_{1}, \Delta_{2}\right]^{T}=\left[\begin{array}{l}
M_{u 1, v 1} \\
M_{u 2, v 2}
\end{array}\right]
$$

(a) If $v 1 \neq v 2$. On the left side we get a non-singular submatrix $\left[2^{v 1-1}, 2^{2(v 1-1)}\right.$; $\left.2^{v 2-1}, 2^{2(v 2-1)}\right]$. So by this we have at most $\sum_{i=1}^{q} \sum_{u 1=1}^{q} \sum_{v 1=1}^{\ell_{u 1}} \sum_{u 2=1}^{q} \sum_{v 2=1}^{\ell_{u 2}}$ $\left(\left(2^{n}-2\right)!\right)$ permutations.
(b) Else if $v 1=v 2=v \in\left[\min \left\{\ell_{u 1}, \ell_{u 2}\right\}\right]$ and $\nexists Y_{i, l 1}=Y_{i, l 2}$ for any distinct $l 1, l 2 \in\left[\ell_{i}\right]$. We get a non-singular submatrix $\left[1,2^{v-1} ; 2^{1}, 2^{v-1}\right]$. By this we have at most $\sum_{i=1}^{q} \sum_{l=1}^{\ell_{i}} \sum_{u 1=1}^{q} \sum_{u 2=1}^{q} \sum_{v=1}^{\min \left\{\ell_{u 1}, \ell_{u 2}\right\}}\left(\left(2^{n}-2\right)!\right)$ permutations.
(c) Else $v 1=v 2=v \in\left[\min \left\{\ell_{u 1}, \ell_{u 2}\right\}\right]$, and $\exists Y_{i, l 1}=Y_{i, l 2}$ for distinct $l 1, l 2 \in$ $\left[\ell_{i}\right]$, we get a non-singular submatrix $\left[0^{n}, 2^{v-1} ; 2^{\ell_{i}-l 1+1} \oplus 2^{\ell_{i}-l 2+1}, 2^{v-1}\right]$, by combing the columns for $Y_{i, l 1}$ and $Y_{i, l 2}$. So by this we have at most $\sum_{i=1}^{q}\binom{\ell_{i}}{2} \sum_{u 1=1}^{q} \sum_{u 2=1}^{q} \sum_{v=1}^{\min \left\{\ell_{u 1}, \ell_{u 2}\right\}}\left(\left(2^{n}-2\right)!\right)$ permutations.
Totally, we have at most $\varsigma_{4}=\left(2 q^{3} \ell^{2}+q^{3} \ell^{3} / 2\right)\left(\left(2^{n}-2\right)!\right)$ permutations can induce this.
5. $\exists \Sigma_{i}=$ Cst $_{j}$ for some $j \in[2]$ and $\Theta_{i}=\Theta_{u}$ for some $u \neq i$. This implies

$$
\left[\begin{array}{cccc}
1, & 1, & \cdots, & 1,  \tag{37}\\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, & \cdots, & 0, \\
2^{1} & 2^{\ell_{u}}, 2^{\ell_{u}-1}, & \cdots, & 0 \\
2^{1}
\end{array}\right] \times \overrightarrow{Y[i, u]}=\left[\begin{array}{c}
\mathrm{Cst}_{j} \\
0^{n}
\end{array}\right],
$$

where $\overrightarrow{Y[i, u]}=\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{u, 1}, Y_{u, 2}, \cdots, Y_{u, \ell_{u}}\right]^{T}, \operatorname{Set} Y[i, u]=$ $\left\{Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{u, 1}, Y_{u, 2}, \cdots, Y_{u, \ell_{u}}\right\}$.
(a) If $\nexists Y_{l 1}^{\prime}, Y_{l 2}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ with any distinct $l 1, l 2 \in\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$.
i. If $\ell_{i}=\ell_{u}$, notice that $M_{i} \neq M_{u}$, so $\exists l \in\left[\ell_{i}\right]$ s.t. $M_{i, l} \neq M_{u, l}$. Then, we get a non-singular submatrix $\left[1,0 ; 2^{\ell_{i}-l+1}, 2^{\ell_{u}-l+1}\right]$.
ii. Else if $\ell_{i}=\ell_{u}+1$, then we focus on the coefficients of $Y_{i, \ell_{i}}, Y_{i, \ell_{i}-1}$ and $Y_{u, \ell_{u}}$, and get a non-singular submatrix $\left[1,1 ; 2^{1}, 2^{2} \oplus 2^{1}\right]$ (when $Y_{i, \ell_{i}-1}=Y_{u, \ell_{u}}$ is a trivial collision) or $\left[1,1 ; 2^{1}, 2^{2}\right]$ (when $Y_{i, \ell_{i}-1} \neq$ $\left.Y_{u, \ell_{u}}\right)$.
iii. Else if $\ell_{i} \geq \ell_{u}+2$, let us focus on the coefficients of $Y_{i, \ell_{i}}$ and $Y_{i, \ell_{i}-1}$, and we get a non-singular submatrix $\left[1,1 ; 2^{2}, 2^{1}\right]$.
iv. Else $\ell_{u} \geq \ell_{i}+1$, let us focus on the coefficients of $Y_{u, \ell_{u}}, Y_{u, \ell_{u}-1}$ and $Y_{i, \ell_{i}}$, and get a non-singular submatrix $\left[1,0 ; 2^{1} \oplus 2^{2}, 2^{1}\right]$ (when $Y_{u, \ell_{u}-1}=Y_{i, \ell_{i}}$ is a trivial collision) or $\left[1,0 ; 2^{1}, 2^{1}\right]$ (when $Y_{u, \ell_{u}-1} \neq$ $\left.Y_{i, \ell_{i}}\right)$.
To summarize this subcase, each case in the above presents us a nonsingular coefficients matrix on the left side, and by this we get at most $\sum_{i=1}^{q} \sum_{u=1, u \neq i}^{q} \sum_{j=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations.
(b) Else $\exists Y_{l 1}^{\prime}, Y_{l 2}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ with distinct $l 1, l 2 \in\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$.
i. If $\ell_{i} \neq \ell_{u}$, then $\bigoplus_{l=1}^{\ell_{i}} 2^{l} \oplus \bigoplus_{l=1}^{\ell_{u}} 2^{l} \neq 0$. On one side, the 2 -collision $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ implies $\left(2^{l 1-1} \oplus 2^{l 2-1}\right) \Delta_{1} \oplus\left(2^{2(l 1-1)} \oplus 2^{2(l 2-1)}\right) \Delta_{2}=M_{l 1}^{\prime} \oplus$ $M_{l 2}^{\prime \prime}$, which is over $\Delta_{1}$ and $\Delta_{2}$. On the other side, some coefficients of Eq. (37) should be combined, if their corresponding variables belong to 2 -collisions or trivial collisions. This makes the final coefficients matrix of Eq. (37) complex. However, notice in this final coefficients matrix that, there is at least one element in its second row should not be 0 , otherwise the sum of all coefficients in the second row should be 0 , and this contradicts with the fact that $\bigoplus_{l=1}^{\ell_{i}} 2^{l} \oplus \bigoplus_{l=1}^{\ell_{u}} 1=$ $\left(2^{1} \oplus 2^{\ell_{i}+1}\right) / 3$ or $\left(2^{1} \oplus 2^{\ell_{i}+1}\right) / 3 \oplus 1$, neither of which is 0 when $1 \leq \ell_{i} \leq 2^{2 n / 3}$. By this we get an equation over $Y_{i, l}$ with $l \in\left[\ell_{i}\right]$, whose coefficient is not 0 . Then, according to the independence of $Y_{i, l}, \Delta_{1}$ and $\Delta_{2}$, we have two independent equations and get at most $\sum_{i=1}^{q} \sum_{u=1, u \neq i}^{q}\binom{\max \left\{\ell_{i}, \ell_{u}\right\}}{2} \sum_{j=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations.
ii. Else $\ell_{i}=\ell_{u}$, on one side by the 2-collision $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ we have an equation over $\Delta_{1}$ and $\Delta_{2}$. On the other side, let us find another equation independent of $\Delta_{1}$ and $\Delta_{2}$. Notice that $M_{i} \neq M_{u}$, so $\exists l \in$ $\left[\ell_{i}\right]$ s.t. $M_{i, l} \neq M_{u, l}$, and this implies $X_{i, l} \neq X_{u, l}$ and $Y_{i, l} \neq Y_{u, l}$. For $Y_{l^{\prime}}^{\prime} \in \operatorname{Set} Y[i, u] \backslash\left\{Y_{i, l}\right\}$, if $\nexists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, then we get an equation over $Y_{i, l}$, whose coefficient is $2^{\ell_{i}-l+1} \neq 0$. Else $\exists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, obviously we have $l^{\prime} \neq l$. Then we get an equation over $Y_{i, l}$, whose coefficient
is either $2^{\ell_{i}-l+1} \oplus 2^{\ell_{i}-l^{\prime}+1} \neq 0\left(\right.$ when $\left.Y_{i, l^{\prime}} \neq Y_{u, l^{\prime}}\right)$ or $2^{\ell_{i}-l+1} \neq 0$ $\left(\right.$ when $\left.Y_{i, l^{\prime}}=Y_{u, l^{\prime}}\right)$.
To summarize this subcase, we get at most $\sum_{i=1}^{q} \sum_{u=1, u \neq i}^{q}$
$\binom{\max \left\{\ell_{i}, \ell_{u}\right\}}{2} \sum_{j=1}^{2}\left(\left(2^{n}-2\right)!\right)$ permutations.
To summarize Case 5 , we get at most $\varsigma_{5}=\left(2 q^{2}+q^{2} \ell^{2}\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
6. $\exists \Sigma_{i}=\Sigma_{u}$ for some $u \neq i$ and $\Theta_{i}=\operatorname{Cst}_{j}$ for some $j \in[2]$. This implies

$$
\left[\begin{array}{ccc}
1, & 1, & \cdots, 1,1,1, \cdots, 1 \\
2^{\ell_{i}}, 2^{\ell_{i}-1}, & \cdots, 2^{1}, 0,0, \cdots, 0
\end{array}\right] \times \overrightarrow{Y[i, u]}=\left[\begin{array}{c}
0^{n} \\
\operatorname{Cst}_{j}
\end{array}\right]
$$

(a) If $\nexists Y_{l 1}^{\prime}, Y_{l 2}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ with any distinct $l 1, l 2 \in\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$.
i. If $\ell_{i}=\ell_{u}$, notice that $M_{i} \neq M_{u}$, so $\exists l \in\left[\ell_{i}\right]$ s.t. $M_{i, l} \neq M_{u, l}$. Then, we get a non-singular submatrix $\left[1,1 ; 2^{\ell_{i}-l+1}, 0\right]$.
ii. Else if $\ell_{i}=\ell_{u}+1$, then we focus on the coefficients of $Y_{i, \ell_{i}}, Y_{i, \ell_{i}-1}$ and $Y_{u, \ell_{u}}$, and get a non-singular submatrix $\left[0,1 ; 2^{2}, 2^{1}\right]$ (when $Y_{i, \ell_{i}-1}=$ $Y_{u, \ell_{u}}$ is a trivial collision) or $\left[1,1 ; 2^{2}, 2^{1}\right]$ (when $Y_{i, \ell_{i}-1} \neq Y_{u, \ell_{u}}$ ).
iii. Else if $\ell_{i} \geq \ell_{u}+2$, let us focus on the coefficients of $Y_{i, \ell_{i}}$ and $Y_{i, \ell_{i}-1}$, and we get a non-singular submatrix $\left[1,1 ; 2^{2}, 2^{1}\right]$.
iv. Else $\ell_{u} \geq \ell_{i}+1$, let us focus on the coefficients of $Y_{u, \ell_{u}}, Y_{u, \ell_{u}-1}$ and $Y_{i, \ell_{i}}$, and get a non-singular submatrix $\left[1,0 ; 0,2^{1}\right]$ (when $Y_{u, \ell_{u}-1}=$ $Y_{i, \ell_{i}}$ is a trivial collision) or $\left[1,1 ; 0,2^{1}\right]$ (when $Y_{u, \ell_{u}-1} \neq Y_{i, \ell_{i}}$ ).
Each case in the above presents us a non-singular coefficients matrix on the left side, and by this we can get at most $\sum_{i=1}^{q} \sum_{u=1, u \neq i}^{q} \sum_{j=1}^{2}$ $\left(\left(2^{n}-2\right)!\right)$ permutations.
(b) Else $\exists Y_{l 1}^{\prime}, Y_{l 2}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l 1}^{\prime}=Y_{l 2}^{\prime \prime}$ with distinct $l 1, l 2 \in\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$. Notice that $\bigoplus_{l=1}^{\ell_{i}} 2^{l} \neq 0$, and then the analysis is similar with Case (5.bi). To summarize, we can get at most $\sum_{i=1}^{q} \sum_{u=1, u \neq i}^{q}\left(\underset{2}{\max \left\{\ell_{i}, \ell_{u}\right\}}\right) \sum_{j=1}^{2}$ ((2 $\left.\left.2^{n}-2\right)!\right)$ permutations.
To summarize Case 6 , we can get at most $\varsigma_{6}=\left(2 q^{2}+q^{2} \ell^{2}\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
7. $\exists \Sigma_{i}=X_{u, v}$ for some $u \in[q], v \in\left[\ell_{u}\right]$ and $\Theta_{i}=\Theta_{j}$ for some $j \neq i$. This implies

$$
\left[\begin{array}{cccccc}
1, & 1, & \cdots, & 1, & 0, & 0, \\
\cdots, & 0, & 2^{v-1}, & 2^{2(v-1)} \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, & \cdots, & 2^{1}, & 2^{\ell_{j}}, 2^{\ell_{j}-1}, & \cdots, \\
2^{1} & 0, & 0
\end{array}\right] \times \overrightarrow{Y[i, j, \Delta]}=\left[\begin{array}{c}
M_{u, v} \\
0^{n}
\end{array}\right]
$$

where $\overrightarrow{Y[i, j, \Delta]}=\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{j, 1}, Y_{j, 2}, \cdots, Y_{j, \ell_{j}}, \Delta_{1}, \Delta_{2}\right]^{T}$, The analysis is similar with that in Case 5 , and their only difference is that, here we have two more variables $\Delta_{1}$ and $\Delta_{2}$. Specially, their coefficients matrix is exactly the same, except for the coefficients for $\Delta_{1}$ and $\Delta_{2}$. Then, we can apply the same analysis, and we can either get a non-singular submatrix on the left side, or get one equation over $\Delta_{1}$ and $\Delta_{2}$, and another equation over $Y_{i, l}, \Delta_{1}$ and $\Delta_{2}$. Finally, in this case we can get at most $\varsigma_{7}=\left(q^{3} \ell+q^{3} \ell^{3} / 2\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
8. $\exists \Sigma_{i}=\Sigma_{j}$ for some $j \neq i$ and $\Theta_{i}=X_{u, v}$ for some $u \in[q], v \in\left[\ell_{u}\right]$. This implies

$$
\left[\begin{array}{ccc}
1, & 1, & \cdots, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \cdots, & 1,1, \cdots, 1, \\
2^{1}, 0,0, \cdots, & 0,2^{v-1}, 2^{2(v-1)}
\end{array}\right] \times \overrightarrow{Y[i, j, \Delta]}=\left[\begin{array}{c}
0^{n} \\
M_{u, v}
\end{array}\right] .
$$

The analysis is similar with that in Case 7, and in this case we get at most $\varsigma_{8}=\left(q^{3} \ell+q^{3} \ell^{3} / 2\right)\left(\left(2^{n}-2\right)!\right)$ permutations.
9. $\Sigma_{i}=\Sigma_{j}$ for some $j \neq i$ and $\Theta_{i}=\Theta_{u}$ for some $u \neq i$, and we have

$$
\left[\begin{array}{cccc}
1, & 1, & \cdots, 1,1,1, \cdots, 1,0, & 0, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \cdots, & \cdots, 2^{1}, 0,0, \cdots, 0,2^{\ell_{u}}, 2^{\ell_{u}-1}, \cdots, 2^{1}
\end{array}\right] \times \overrightarrow{Y[i, j, u]}=\left[\begin{array}{c}
0^{n} \\
0^{n}
\end{array}\right],
$$

where $\overrightarrow{Y[i, j, u]}=\left[Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{j, 1}, Y_{j, 2}, \cdots, Y_{j, \ell_{j}}, Y_{u, 1}, Y_{u, 2}, \cdots, Y_{u, \ell_{u}}\right]^{T}$.
(a) If $j=u \wedge \nexists Y_{l^{\prime}}^{\prime}, Y_{l^{\prime \prime}}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l^{\prime}}^{\prime}=Y_{l^{\prime \prime}}^{\prime \prime}$ with any distinct $l^{\prime}, l^{\prime \prime} \in$ $\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$. The equation set turns to be

$$
\left[\begin{array}{cccc}
1, & 1, & \cdots, & 1,  \tag{38}\\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, & \cdots, & 2^{1}, \\
2^{\ell_{u}} & 2^{\ell_{u}-1}, & \cdots, & \cdots, \\
2^{1}
\end{array}\right] \times \overrightarrow{Y[i, u]}=\left[\begin{array}{c}
0^{n} \\
0^{n}
\end{array}\right] .
$$

i. If $\ell_{i}=\ell_{u}$, let us denote $Y_{*, l}=Y_{i, l} \oplus Y_{u, l}$, then Eq. (38) becomes

$$
\left[\begin{array}{cc}
1, & 1, \\
2^{\ell_{i}}, & 2^{\ell_{i}-1}, \cdots, \\
, & 2^{1}
\end{array}\right] \times\left[Y_{*, 1}, Y_{*, 2}, \cdots, Y_{*, \ell_{i}}\right]^{T}=\left[\begin{array}{c}
0^{n} \\
0^{n}
\end{array}\right] .
$$

On the left side the coefficients matrix is an MDS matrix, and on the right side we have two $0^{n}$, so by the property of MDS matrix and the fact $M_{i} \neq M_{u}$, we have at least 3 non-zero $Y_{*, l}$. This means in Eq. (38) we have distinct $l 1, l 2, l 3 \in\left[\ell_{i}\right]$ s.t. $Y_{i, l 1} \neq Y_{u, l 1}, Y_{i, l 2} \neq Y_{u, l 2}$ and $Y_{i, l 3} \neq Y_{u, l 3}$. Then in Eq. (38) we find a non-singular submatrix $\left[1,1 ; 2^{\ell_{i}-l 1+1}, 2^{\ell_{i}-l 2+1}\right]$.
ii. Else if $\ell_{i}=\ell_{u}+1$, then we focus on the coefficients of $Y_{i, \ell_{i}}, Y_{i, \ell_{i}-1}$ and $Y_{u, \ell_{u}}$, and get a non-singular submatrix $\left[1,0 ; 2^{1}, 2^{2} \oplus 2^{1}\right]$ (when $Y_{i, \ell_{i}-1}=Y_{u, \ell_{u}}$ is a trivial collision) or $\left[1,1 ; 2^{1}, 2^{2}\right]$ (when $Y_{i, \ell_{i}-1} \neq$ $\left.Y_{u, \ell_{u}}\right)$.
iii. Else if $\ell_{i} \geq \ell_{u}+2$, let us focus on the coefficients of $Y_{i, \ell_{i}}$ and $Y_{i, \ell_{i}-1}$, and we get a non-singular submatrix $\left[1,1 ; 2^{2}, 2^{1}\right]$.
iv. Else $\ell_{u} \geq \ell_{i}+1$, the analysis is the same as (ii) and (iii).
(b) Else if $j=u \wedge \exists Y_{l^{\prime}}^{\prime}, Y_{l^{\prime \prime}}^{\prime \prime} \in \operatorname{Set} Y[i, u]$ s.t. $Y_{l^{\prime}}^{\prime}=Y_{l^{\prime \prime}}^{\prime \prime}$ with distinct $l^{\prime}, l^{\prime \prime} \in$ $\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$. On one side by the 2 -collision $Y_{l^{\prime}}^{\prime}=Y_{l^{\prime \prime}}^{\prime \prime}$ we have an equation over $\Delta_{1}$ and $\Delta_{2}$. On the other side, let us find another equation independent of $\Delta_{1}$ and $\Delta_{2}$.
i. If $\ell_{i} \neq \ell_{u}$, then $\bigoplus_{l=1}^{\ell_{i}} 2^{l} \oplus \bigoplus_{l=1}^{\ell_{u}} 2^{l} \neq 0$, we get an equation over $Y_{i, l}$, and the analysis is similar with (5.b).
ii. Else $\ell_{i}=\ell_{u}$, notice that $M_{i} \neq M_{u}$, so $\exists l \in\left[\ell_{i}\right]$ s.t. $M_{i, l} \neq M_{u, l}$, and this implies $Y_{i, l} \neq Y_{u, l}$. For $Y_{l^{\prime}}^{\prime} \in \operatorname{Set} Y[i, u] \backslash\left\{Y_{i, l}\right\}$, if $\nexists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, then we get an equation over $Y_{i, l}$, whose coefficient is $2^{\ell_{i}-l+1} \neq 0$.

Else $\exists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, obviously we have $l^{\prime} \neq l$. Then we get an equation over $Y_{i, l}$, whose coefficient is either $2^{\ell_{i}-l+1} \oplus 2^{\ell_{i}-l^{\prime}+1} \neq 0$ (when $\left.Y_{i, l^{\prime}} \neq Y_{u, l^{\prime}}\right)$ or $2^{\ell_{i}-l+1} \neq 0\left(\right.$ when $\left.Y_{i, l^{\prime}}=Y_{u, l^{\prime}}\right)$.
(c) Else if $j \neq u \wedge \nexists Y_{l^{\prime}}^{\prime}, Y_{l^{\prime \prime}}^{\prime \prime} \in \operatorname{Set} Y[i, j, u]$ s.t. $Y_{l^{\prime}}^{\prime} \stackrel{=}{=} Y_{l^{\prime \prime}}^{\prime \prime}$ with any distinct $l^{\prime}, l^{\prime \prime} \in\left[\max \left\{\ell_{i}, \ell_{j}, \ell_{u}\right\}\right]$. By $M_{i} \neq M_{j}$ we know $\exists l 1 \in\left[\max \left\{\ell_{i}, \ell_{j}\right\}\right]$ s.t. $Y_{i, l 1} \neq Y_{j, l 1}$. Here $Y_{i, l 1}=0^{n}$ if $l 1>\ell_{i}$ and $Y_{j, l 1}=0^{n}$ if $l 1>\ell_{j}$. Notice that $Y_{i, l 1} \oplus Y_{j, l 1} \neq 0^{n}$ can be seen as a new variable. We ignore $Y_{u, l 1}$ here because its coefficient is $0^{n}$ in the first row of coefficients matrix. Similarly, by $M_{i} \neq M_{u}$ we know $\exists l 2 \in\left[\max \left\{\ell_{i}, \ell_{u}\right\}\right]$ s.t. $Y_{i, l 2} \neq Y_{u, l 2}$. Here $Y_{i, l 2}=0^{n}$ if $l 2>\ell_{i}$ and $Y_{u, l 2}=0^{n}$ if $l 2>\ell_{u}$. Then $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus$ $Y_{u, l 2}$ is a new variable. We ignore $Y_{j, l 2}$ here because its coefficient is $0^{n}$ in the second row of coefficients matrix. Also by $M_{j} \neq M_{u}$ we have $\exists l 3 \in\left[\max \left\{\ell_{j}, \ell_{u}\right\}\right]$ s.t. $Y_{j, l 3} \neq Y_{u, l 3}$. Here $Y_{j, l 3}=0^{n}$ if $l 3>\ell_{j}$ and $Y_{u, l 3}=0^{n}$ if $l 3>\ell_{u}$.
If $l 1 \neq l 2$, it is easy to see that $Y_{i, l 1} \oplus Y_{j, l 1}$ and $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus 2^{\ell_{u}-l 2+1} Y_{u, l 2}$ are independent of each other, because we have $Y_{l^{\prime}}^{\prime} \neq Y_{l^{\prime \prime}}^{\prime \prime}$ with any distinct $l^{\prime}, l^{\prime \prime} \in\left[\max \left\{\ell_{i}, \ell_{j}, \ell_{u}\right\}\right]$. If $l 1=l 2$ and $Y_{j, l 1} \neq Y_{u, l 2}$, then $Y_{i, l 1} \oplus$ $Y_{j, l 1}$ and $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus 2^{\ell_{u}-l 2+1} Y_{u, l 2}$ are also independent. If $l 1=l 2$ and $Y_{j, l 1}=Y_{u, l 2}$, notice that $Y_{j, l 3} \neq Y_{u, l 3}$ and $l 2 \neq l 3$, we have variables $Y_{i, l 1} \oplus Y_{j, l 1} \oplus Y_{j, l 3}$ and $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus 2^{\ell_{u}-l 2+1} Y_{u, l 2} \oplus 2^{\ell_{u}-l 3+1} Y_{u, l 3}$ are independent.
Then we find a non-singular submatrix in the above coefficients matrix, i.e. $[1,0 ; 0,1]$ for independent variables $Y_{i, l 1} \oplus Y_{j, l 1}$ and $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus$ $2^{\ell_{u}-l 2+1} Y_{u, l 2}$ or $Y_{i, l 1} \oplus Y_{j, l 1} \oplus Y_{j, l 3}$ and $2^{\ell_{i}-l 2+1} Y_{i, l 2} \oplus 2^{\ell_{u}-l 2+1} Y_{u, l 2} \oplus$ $2^{\ell_{u}-l 3+1} Y_{u, l 3}$.
(d) Else $j \neq u \wedge \exists Y_{l^{\prime}}^{\prime}, Y_{l^{\prime \prime}}^{\prime \prime} \in \operatorname{Set} Y[i, j, u]$ s.t. $Y_{l^{\prime}}^{\prime}=Y_{l^{\prime \prime}}^{\prime \prime}$ with distinct $l^{\prime}, l^{\prime \prime} \in$ $\left[\max \left\{\ell_{i}, \ell_{j}, \ell_{u}\right\}\right]$. On one side by the 2 -collision $Y_{l^{\prime}}^{\prime}=Y_{l^{\prime \prime}}^{\prime \prime}$ we have an equation over $\Delta_{1}$ and $\Delta_{2}$. On the other side, let us find another equation independent of $\Delta_{1}$ and $\Delta_{2}$.
i. If $\ell_{i} \neq \ell_{u}$, we have $\bigoplus_{l=1}^{\ell_{i}} 2^{l} \oplus \bigoplus_{l=1}^{\ell_{u}} 2^{l} \neq 0$, so we get an equation over $Y_{i, l}$ for some $l \in\left[\ell_{i}\right]$.
ii. Else $\ell_{i}=\ell_{u}$, notice that $M_{i} \neq M_{u}$, so $\exists l \in\left[\ell_{i}\right]$ s.t. $M_{i, l} \neq M_{u, l}$, and this implies $Y_{i, l} \neq Y_{u, l}$. For $Y_{l^{\prime}}^{\prime} \in \operatorname{Set} Y[i, j, u] \backslash\left\{Y_{i, l}\right\}$, if $\nexists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, then we get an equation over $Y_{i, l}$, whose coefficient is $2^{\ell_{i}-l+1} \neq 0$. Else $\exists Y_{l^{\prime}}^{\prime}=Y_{i, l}$, let us focus on the second row of the coefficients matrix. By this we can ignore the influence from $M_{j}$, then we have $Y_{l^{\prime}}^{\prime}=Y_{i, l^{\prime}}$ or $Y_{l^{\prime}}^{\prime}=Y_{u, l^{\prime}}$, so it is obvious that $l^{\prime} \neq l$. Then we get an equation over $Y_{i, l}$, whose coefficient is either $2^{\ell_{i}-l+1} \oplus 2^{\ell_{i}-l^{\prime}+1} \neq 0$ (when $Y_{i, l^{\prime}} \neq Y_{u, l^{\prime}}$ ) or $2^{\ell_{i}-l+1} \neq 0$ (when $Y_{i, l^{\prime}}=Y_{u, l^{\prime}}$ ).
To summarize, we can get at most $\varsigma_{9}=\left(\sum_{i=1}^{q} \sum_{j=1, j \neq i}^{q}\left(1+\binom{\max \left\{\ell_{i}, \ell_{j}\right\}}{2}\right)+\right.$ $\left.\sum_{i=1}^{q} \sum_{j=1, j \neq i}^{q} \sum_{u=1, u \neq i, j}^{q}\left(1+\left(\underset{2}{\max \left\{\ell_{i}, \ell_{j}, \ell_{u}\right\}}\right)\right)\right)\left(\left(2^{n}-2\right)!\right) \leq\left(q^{2}+q^{2} \ell_{\max }^{2} / 2+\right.$ $\left.q^{3}+q^{3} \ell_{\text {max }}^{2} / 2\right)\left(\left(2^{n}-2\right)!\right.$ ! permutations.

Finally, we can get

$$
\epsilon_{e c f} \leq \frac{\sum_{i=-1}^{9} \varsigma_{i}}{2^{n}!} \leq \frac{3 q \ell\left(\left(2^{n}-1\right)!\right)+20 q^{3} \ell^{3}\left(\left(2^{n}-2\right)!\right)}{2^{n}!} \leq \frac{3 q \ell}{2^{n}}+\frac{5 q^{3} \ell^{3}}{2^{2 n-3}}
$$

### 9.3 Upper Bounding pseudo-cover-free $\boldsymbol{\epsilon}_{p c f}$

According to the definition of $\epsilon_{p c f}$, we need to upper bound the occurrence probability of 36 bad events, as listed in Table. 5 and 6.

1. For the cases from 1 to $6, \Sigma_{i}=\operatorname{Cst}_{j 1}$ for some $j 1 \in[2]$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} Y_{i, l}=\mathrm{Cst}_{j 1}$, i.e. an equation over variables $Y_{i, l}$.
On the other side, notice that by $\Sigma_{i}=\mathrm{Cst}_{j 1}$ we have $\pi\left(\Sigma_{i}\right)=\pi\left(\mathrm{Cst}_{j 1}\right)=$ $\Delta_{j 1}$. This means, $\pi\left(\Sigma_{i}\right) \oplus c_{i}=\Delta_{j 1} \oplus c_{i}$. Then, though the cases from 1 to 6 imply different equations, they all depend on random variable $\Delta_{j 1}$. Notice that we have restrict $c_{i} \neq 0^{n}$ for all $i \in[q]$, so the equation from case 6 can't trivially hold.
Then, we find two equations independent of each other, and by this we have at most $\sum_{i=1}^{6} \varsigma_{i} \leq 12 q^{2} \ell\left(\left(2^{n}-2\right)!\right)$ permutations.
2. For the cases from 7 to $12, \Sigma_{i}=X_{u, v}$ for some $u \neq i, v \in\left[\ell_{j 1}\right]$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} Y_{i, l}=2^{v-1} \Delta_{1} \oplus 2^{2(v-2)} \Delta_{2} \oplus M_{v}^{u}$, i.e. an equation over variables $Y_{i, l}$, $\Delta_{1}$ and $\Delta_{2}$.
(a) For case $7, \pi\left(\Sigma_{i}\right) \oplus c_{i}=\Delta_{j 2}$, we get a non-singular submatrix $[1,1 ; 1,0]$ or $[1,1 ; 0,1]$ with variables $\Delta_{1}$ and $\Delta_{2}$.
(b) For case $8, \pi\left(\Sigma_{i}\right) \oplus c_{i}=Y_{j 2, l 2}$ depends only on $Y_{j 2, l 2}$, where $c_{i} \neq 0^{n}$ excludes trivial collisions $Y_{j 2, l 2}=Y_{l 2}^{\prime}$. Then we get two independent equations.
(c) For cases from 9 to 12 , we need only to consider $\pi\left(\Sigma_{i}\right) \oplus c_{i}=c_{i^{\prime}, j^{\prime}}$ with $c_{i^{\prime}, j^{\prime}} \notin\left\{c_{i, 1}, c_{i, 2}\right\}$ and $c_{i^{\prime}, j^{\prime}} \notin\left\{\Delta_{u}, Y_{k, l}\right\}$. Such $c_{i^{\prime}, j^{\prime}}$ are produced by $c_{i^{\prime}, j^{\prime}}=\pi\left(\Sigma_{i^{\prime}, j^{\prime}}\right)$ with $\Sigma_{i^{\prime}, j^{\prime}} \notin\left\{\mathrm{Cst}_{u}, X_{k, l}\right\}$ or $c_{i^{\prime}, j^{\prime}}=\pi\left(\Theta_{i^{\prime}, j^{\prime}}\right)$ with $\Theta_{i^{\prime}, j^{\prime}} \notin\left\{\operatorname{Cst}_{u}, X_{k, l}\right\}$ (otherwise such cases should have been analyzed in (a) and (b)), so they have independent randomness from $\left\{\Delta_{u}, Y_{k, l}\right\}$.

Then, by two independent equations we have at most $\sum_{i=7}^{12} \varsigma_{i} \leq 6 q^{3} \ell^{2}\left(\left(2^{n}-\right.\right.$ 2)!) permutations.
3. For cases from 13 to $18, \Sigma_{i}=\Sigma_{j}$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} Y_{i, l}=\bigoplus_{l=1}^{\ell_{j}} Y_{j, l}$, depending only on $Y_{i, l}$ and $Y_{j, l}$. Also, we have $c_{i, 1}=\pi\left(\Sigma_{i}\right)=\pi\left(\Sigma_{j}\right)=c_{j, 1}$, which implies $\pi\left(\Sigma_{i}\right) \oplus c_{i}=c_{j, 1} \oplus c_{i}$.
(a) For case $13, c_{j, 1} \oplus c_{i}=\Delta_{j 2}$ depending on $\Delta_{1}$ or $\Delta_{2}$, and this always holds regardless of whether $c_{j, 1} \in\left\{\operatorname{Cst}_{u}, X_{k, l}\right\}$ or not.
(b) For case 14, $c_{j, 1} \oplus c_{i}=Y_{u, v}$ for some $u \in[q]$ and $v \in\left[L_{u}\right]$. When $c_{j, 1} \in$ $\left\{\Delta_{1}, \Delta_{2}\right\}$ or $c_{j, 1}=P\left(\Sigma_{j}\right)$ with $\Sigma_{j} \notin\left\{\mathrm{Cst}_{u}, X_{k, l}\right\}$, the two equations are independent; when $c_{j, 1} \in\left\{Y_{k, l}\right\}$, say $c_{j, 1}=Y_{u^{\prime}, v^{\prime}}$, then $c_{i} \neq 0^{n}$ excludes trivial collisions and so $Y_{u^{\prime}, v^{\prime}}$ and $Y_{u, v}$ are independent, both with coefficients 1. Notice in the first equation that we have at least one $Y_{i, l 1}$ on its left side, with coefficient 1. If $Y_{i, l 1} \notin\left\{Y_{u^{\prime}, v^{\prime}}, Y_{u, v}\right\}$, the two equations are independent. If $Y_{i, l 1} \in\left\{Y_{u^{\prime}, v^{\prime}}, Y_{u, v}\right\}$, we can get a non-singular submatrix $[1,0 ; 1,1]$. It is possible that $\left\{Y_{i, l 1}, Y_{j, l 1}\right\}=\left\{Y_{u^{\prime}, v^{\prime}}, Y_{u, v}\right\}$, which results in a singular submatrix $[1,1 ; 1,1]$. In such a case, notice that there must be some other $Y_{i, l}$ or $Y_{u, l}$ in the first equation, otherwise we get a contradiction $Y_{i, l 1}=Y_{j, l 1}$. This helps the first equation to be independent of the second one. If there exist 2-collisions among
$\left\{Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{j, 1}, Y_{j, 2}, \cdots, Y_{j, \ell_{j}}\right\}$ in the first equation, then itself implies two independent equations.
(c) For cases from 15 to $18, c_{j, 1} \oplus c_{i} \notin\left\{\Delta_{1}, \Delta_{1}, Y_{k, l}\right\}$, then it has independent randomness from the first equation.
At last, by noticing that in each subcase we have two independent equations, here we can get at most $\sum_{i=13}^{18} \varsigma_{i} \leq 6 q^{3} \ell\left(\left(2^{n}-2\right)!\right.$ ) permutations.

1. For the cases from 19 to $24, \Theta_{i}=\mathrm{Cst}_{j 1}$ for some $j 1 \in[2]$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} 2^{\ell_{i}-l+1} Y_{i, l}=$ Cst $_{j 1}$, i.e. an equation over variables $Y_{i, l}$.
On the other side, notice that by $\Theta_{i}=\operatorname{Cst}_{j 1}$ we have $\pi\left(\Theta_{i}\right)=\pi\left(\operatorname{Cst}_{j 1}\right)=$ $\Delta_{j 1}$. This means, $\pi\left(\Theta_{i}\right) \oplus c_{i}=\Delta_{j 1} \oplus c_{i}$. Then, though the cases from 1 to 6 imply different equations, they all depend on random variable $\Delta_{j 1}$. Notice that we have restrict $c_{i} \neq 0^{n}$ for all $i \in[q]$, so the equation from case 6 can't trivially hold.
Then, we find two equations independent of each other, and by this we have at most $\sum_{i=19}^{24} \varsigma_{i} \leq 12 q^{2} \ell\left(\left(2^{n}-2\right)!\right)$ permutations.
2. For the cases from 25 to $30, \Theta_{i}=X_{u, v}$ for some $u \neq i, v \in\left[\ell_{j 1}\right]$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} 2^{\ell_{i}-l+1} Y_{i, l}=2^{v-1} \Delta_{1} \oplus 2^{2(v-2)} \Delta_{2} \oplus M_{v}^{u}$, i.e. an equation over variables $Y_{i, l}, \Delta_{1}$ and $\Delta_{2}$.
(a) For case $25, \pi\left(\Theta_{i}\right) \oplus c_{i}=\Delta_{j 2}$, we get a non-singular submatrix $\left[2^{v-1}, 2^{2(v-2)}\right.$; $1,0]$ or $\left[2^{v-1}, 2^{2(v-2)} ; 0,1\right]$ with variables $\Delta_{1}$ and $\Delta_{2}$.
(b) For case $26, \pi\left(\Theta_{i}\right) \oplus c_{i}=Y_{j 2, l 2}$ depends only on $Y_{j 2, l 2}$, where $c_{i} \neq 0^{n}$ excludes trivial collisions $Y_{j 2, l 2}=Y_{l 2}^{\prime}$. Then we get two independent equations.
(c) For cases from 27 to 30 , we need only to consider $\pi\left(\Theta_{i}\right) \oplus c_{i}=c_{i^{\prime}, j^{\prime}}$ with $c_{i^{\prime}, j^{\prime}} \notin\left\{c_{i, 1}, c_{i, 2}\right\}$ and $c_{i^{\prime}, j^{\prime}} \notin\left\{\Delta_{u}, Y_{k, l}\right\}$. Such $c_{i^{\prime}, j^{\prime}}$ are produced by $c_{i^{\prime}, j^{\prime}}=\pi\left(\Sigma_{i^{\prime}, j^{\prime}}\right)$ with $\Sigma_{i^{\prime}, j^{\prime}} \notin\left\{\mathrm{Cst}_{u}, X_{k, l}\right\}$ or $c_{i^{\prime}, j^{\prime}}=\pi\left(\Theta_{i^{\prime}, j^{\prime}}\right)$ with $\Theta_{i^{\prime}, j^{\prime}} \notin\left\{\operatorname{Cst}_{u}, X_{k, l}\right\}$ (otherwise such cases should have been analyzed in (a) and (b)), so they have independent randomness from $\left\{\Delta_{u}, Y_{k, l}\right\}$.

Then, by two independent equations we have at most $\sum_{i=25}^{30} \varsigma_{i} \leq 6 q^{2} \ell^{2}\left(\left(2^{n}-\right.\right.$ 2)!) permutations.
3. For cases from 31 to $36, \Theta_{i}=\Theta_{j}$ is equivalent to $\bigoplus_{l=1}^{\ell_{i}} 2^{\ell_{i}-l+1} Y_{i, l}=$ $\bigoplus_{l=1}^{\ell_{j}} 2^{\ell_{j}-l+1} Y_{j, l}$, depending only on $Y_{i, l}$ and $Y_{j, l}$. Also, we have $c_{i, 2}=$ $\pi\left(\Theta_{i}\right)=\pi\left(\Theta_{j}\right)=c_{j, 2}$, which implies $\pi\left(\Theta_{i}\right) \oplus c_{i}=c_{j, 2} \oplus c_{i}$.
(a) For case $31, c_{j, 2} \oplus c_{i}=\Delta_{j 2}$ depending on $\Delta_{1}$ or $\Delta_{2}$, and this always holds regardless of whether $c_{j, 1} \in\left\{\mathrm{Cst}_{u}, X_{k, l}\right\}$ or not.
(b) For case $32, c_{j, 2} \oplus c_{i}=Y_{u, v}$ for some $u \in[q]$ and $v \in\left[L_{u}\right]$. When $c_{j, 2} \in\left\{\Delta_{1}, \Delta_{2}\right\}$ or $c_{j, 2}=P\left(\Theta_{j}\right)$ with $\Theta_{j} \notin\left\{\operatorname{Cst}_{u}, X_{k, l}\right\}$, the two equations are independent; when $c_{j, 2} \in\left\{Y_{k, l}\right\}$, say $c_{j, 2}=Y_{u^{\prime}, v^{\prime}}$, then $c_{i} \neq 0^{n}$ excludes trivial collisions and so $Y_{u^{\prime}, v^{\prime}}$ and $Y_{u, v}$ are independent, both with coefficients 1 . Notice in the first equation that $M_{i} \neq M_{j}$ implies $\exists l 1 \in\left[\max \left\{\ell_{i}, \ell_{j}\right\}\right]$ s.t. $Y_{i, l 1} \neq Y_{j, l 1}$, where $Y_{i, l 1}=0^{n}$ if $l 1>\ell_{i}$ and $Y_{j, l 1}=0^{n}$ if $l 1>\ell_{j}$. The variable $Y_{i, l 1} \oplus Y_{j, l 1}$ has a coefficient $2^{\max \left\{\ell_{i}, \ell_{j}\right\}-l 1+1} \neq 1$. If $Y_{i, l 1} \oplus Y_{j, l 1} \notin\left\{Y_{u^{\prime}, v^{\prime}}, Y_{u, v}\right\}$, the two equations are
independent. If $Y_{i, l 1} \oplus Y_{j, l 1} \in\left\{Y_{u^{\prime}, v^{\prime}}, Y_{u, v}\right\}$, we can get a non-singular submatrix $\left[2^{\max \left\{\ell_{i}, \ell_{j}\right\}-l 1+1}, 0 ; 1,1\right]$ or $\left[2^{\max \left\{\ell_{i}, \ell_{j}\right\}-l 1+1}, 1 ; 1,1\right]$. If there exist 2-collisions among $\left\{Y_{i, 1}, Y_{i, 2}, \cdots, Y_{i, \ell_{i}}, Y_{j, 1}, Y_{j, 2}, \cdots, Y_{j, \ell_{j}}\right\}$ in the first equation, then itself implies two independent equations.
(c) For cases from 33 to $36, c_{j, 2} \oplus c_{i} \notin\left\{\Delta_{1}, \Delta_{1}, Y_{k, l}\right\}$, then it has independent randomness from the first equation.
By noticing that in each subcase we have two independent equations, here we can get at most $\sum_{i=31}^{36} \varsigma_{i} \leq 6 q^{3} \ell\left(\left(2^{n}-2\right)\right.$ !) permutations.

Finally, we can get

$$
\epsilon_{p c f} \leq \frac{\sum_{i=1}^{36} \varsigma_{i}}{2^{n!}} \leq \frac{24 q^{3} \ell^{2}\left(\left(2^{n}-2\right)!\right)}{2^{n}!} \leq \frac{3 q^{3} \ell^{2}}{2^{2 n-4}}
$$

### 9.4 Upper Bounding block-wise universal $\epsilon_{a u}$

By its definition, we have two bad events in upper bounding $\epsilon_{a u}$.

1. $\Sigma_{i}=\Sigma_{j}$ for some $j \neq i$. This implies an equation

$$
\bigoplus_{l=1}^{\ell_{i}} Y_{i, l}=\bigoplus_{l=1}^{\ell_{j}} Y_{j, l}
$$

Notice that $M_{i} \neq M_{j}$, so there exists $l^{\prime} \in\left[\max \left\{\ell_{i}, \ell_{j}\right\}\right]$ s.t. $Y_{i, l^{\prime}} \neq Y_{j, l^{\prime}}$, where $Y_{i, l^{\prime}}=0^{n}$ if $l^{\prime}>\ell_{i}$ and $Y_{j, l^{\prime}}=0^{n}$ if $l^{\prime}>\ell_{j}$. Then, the variable $Y_{i, l^{\prime}} \oplus Y_{j, l^{\prime}}$ has a non-zero coefficient, so we have

$$
\operatorname{Pr}\left[\Sigma_{i}=\Sigma_{j}\right] \leq \frac{1}{2^{n}-(q \ell-2-2 q)} \leq \frac{1}{2^{n-1}}
$$

2. $\Theta_{i}=\Theta_{j}$ for some $j \neq i$. This implies an equation

$$
\bigoplus_{l=1}^{\ell_{i}} 2^{\ell_{i}-l+1} Y_{i, l}=\bigoplus_{l=1}^{\ell_{j}} 2^{\ell_{j}-l+1} Y_{j, l}
$$

Notice that $M_{i} \neq M_{j}$, so there exists $l^{\prime} \in\left[\max \left\{\ell_{i}, \ell_{j}\right\}\right]$ s.t. $Y_{i, l^{\prime}} \neq Y_{j, l^{\prime}}$, where $Y_{i, l^{\prime}}=0^{n}$ if $l^{\prime}>\ell_{i}$ and $Y_{j, l^{\prime}}=0^{n}$ if $l^{\prime}>\ell_{j}$. Then, the variable $2^{\ell_{i}-l^{\prime}+1} Y_{i, l^{\prime}} \oplus 2^{\ell_{j}-l^{\prime}+1} Y_{j, l^{\prime}}$ has a non-zero coefficient, so we have

$$
\operatorname{Pr}\left[\Sigma_{i}=\Sigma_{j}\right] \leq \frac{1}{2^{n}-(q \ell-2-2 q)} \leq \frac{1}{2^{n-1}}
$$

In conclusion, we have $\epsilon_{a u} \leq 2^{1-n}$.

## 10 Conclusion

With the fast developments of computing power, birthday attacks gradually become practical threats to cryptographic algorithms, and this is especially serious for modes of operation on small-size block ciphers. Compared with the passive ways that just enlarge the sizes of internal states and outputs, designing beyond-birthday-bound schemes is active and promising.

We successfully unify the three independent keys in the current beyond-birthday-bound MAC modes in this paper, by developing several theorems that can reduce the security of three/two/one-key such constructions to some properties on internal structures and PRP assumption on block ciphers. Our developed tools are also useful to simplify the analysis for other modes of operations, which is of independent interests.

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|  | $\mathcal{L}(Y)$ | $\# \operatorname{acc}(\sim)$ | \#choices | Rank of $(\mathcal{L}(y), \sim)$ | Total |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 3-message fully-covered |  | 0 | 2 | 2 |  |
|  | $\Sigma_{i}=\Sigma_{j}, \Theta_{i}=\Theta_{k}$ | 1 | $l^{2}$ | 2 | $\frac{l^{2}}{2^{2 n}}$ |
|  |  | 0 | $l$ | 2 |  |
|  | $\Sigma_{i}=\Sigma_{j}, \Theta_{i}=X_{k}$ | 1 | $l^{3}$ | 2 | $\frac{l^{3}}{2^{2 n}}$ |
|  |  | 0 | $l$ | 2 |  |
|  | $\Sigma_{i}=X_{j}, \Theta_{i}=\Theta_{k}$ | 1 | $l^{3}$ | 2 | $\frac{l^{3}}{2^{2 n}}$ |
|  |  | 0 | $l^{2}$ | 2 |  |
|  | $\Sigma_{i}=X_{j}, \Theta_{i}=X_{k}$ | 1 | $l^{4}$ | 2 | $\frac{l^{4}}{2^{2 n}}$ |
| 2-message single-covered |  | 0 | 2 | 1 |  |
|  | $\Sigma_{i}=\Sigma_{j}$ | 1 | $l^{2}$ | 1 | $\frac{l^{2}}{2^{n}}$ |
|  |  | 0 | $l$ | 1 |  |
|  | $\Sigma_{i}=X_{j}$ | 1 | $l^{3}$ | 1 | $\frac{l^{3}}{2^{n}}$ |
|  |  | 0 | $l$ | 1 |  |
|  | $\Theta_{i}=X_{k}$ | 1 | $l^{3}$ | 1 | $\frac{l^{3}}{2^{n}}$ |
|  |  | 0 | 2 | 1 |  |
|  | $\Theta_{i}=\Theta_{k}$ | 1 | $l^{2}$ | 1 | $\frac{l^{2}}{2^{n}}$ |
| 3-message pseudo-covered |  | 0 | $9 l^{2}$ | 2 |  |
|  | $\Sigma_{i}=X_{j}, Y_{j}+t_{i}=Y_{k}$ | $1$ | $l^{4}$ | 2 | $\frac{l^{4}}{2^{n}}$ |
|  |  | 0 | $9 l^{2}$ | 2 |  |
|  | $\Theta_{i}=X_{j}, Y_{j}+t_{i}=Y_{k}$ | 1 | $l^{4}$ | 2 | $\frac{l^{4}}{2^{n}}$ |
| 4-message pseudo-covered | $\Sigma_{i}=X_{e}, \Sigma_{j}=X_{f}, Y_{e}+Y_{f}=t_{i}+t_{j}$ | 0 | $l^{2}$ | 3 |  |
|  |  | 1 | $l^{4}$ | 3 | $\frac{l^{4}}{2^{3 n}}+\frac{l^{6}}{2^{4 n}}$ |
|  |  | 2 | $l^{6}$ | 4 | $2^{3 n}+\frac{2^{4 n}}{}$ |
|  | $\Sigma_{i}=X_{e}, \Theta_{j}=X_{f}, Y_{e}+Y_{f}=t_{i}+t_{j}$ | 0 | $l^{2}$ | 3 |  |
|  |  | 1 | $l^{4}$ | 3 | $\frac{l^{4}}{2^{3 n}}+\frac{l^{6}}{2^{4 n}}$ |
|  |  | 2 | $l^{6}$ | 4 |  |
|  | $\Theta_{i}=X_{e}, \Sigma_{j}=X_{f}, Y_{e}+Y_{f}=t_{i}+t_{j}$ | 0 | $l^{2}$ | 3 |  |
|  |  | 1 | $l^{4}$ | 3 | $\frac{l^{4}}{2^{3 n}}+\frac{l^{6}}{2^{4 n}}$ |
|  |  | 2 | $l^{6}$ | 4 |  |
|  | $\Theta_{i}=X_{e}, \Theta_{j}=X_{f}, Y_{e}+Y_{f}=t_{i}+t_{j}$ | 0 | $l^{2}$ | 3 |  |
|  |  | 1 | $l^{4}$ | 3 | $\frac{l^{4}}{2^{3 n}}+\frac{l^{6}}{2^{4 n}}$ |
|  |  | 2 | $l^{6}$ | 4 |  |

Table 3. Table for different cases of bad equations with no. of choice and ranks corresponding to accidents 0,1 and 2 .

Table 4. 9 Bad Events in Upper Bounding $\epsilon_{e c f}$.

| $\Sigma_{i}=$ | $\Theta_{i}=$ | $\operatorname{Cst}_{j 2}$ | $X_{j 2, l}$ |
| :--- | :---: | :---: | :---: |
| $\Theta_{j 2}$ |  |  |  |
| Cst $_{j 1}$ | 1 | 2 | 5 |
| $X_{j 1, l}$ | 3 | 4 | 7 |
| $\Sigma_{j 1}$ | 6 | 8 | 9 |

Table 5. 18 out of 36 Bad Events in Upper Bounding $\epsilon_{p c f}$ : 1st Half.

| $\Sigma_{i}=$ | $\pi\left(\Sigma_{i}\right) \oplus c_{i}$ | $\pi\left(\Sigma_{j 2}\right) \oplus c_{j 2}$ | $\pi\left(\Theta_{j 2}\right) \oplus c_{j 2}$ | $\pi\left(\Theta_{j 2}\right)$ | $\pi\left(\Sigma_{j 2}\right)$ | $Y_{j 2, l}$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| $\Delta_{j 2}$ |  |  |  |  |  |  |
| $\mathrm{Cst}_{j 1}$ | 1 | 2 | 3 | 4 | 5 | 6 |
| $X_{j 1, l}$ | 9 | 10 | 11 | 12 | 8 | 7 |
| $\Sigma_{j 1}$ | 15 | 16 | 17 | 18 | 14 | 13 |

Table 6. 18 out of 36 Bad Events in Upper Bounding $\epsilon_{p c f}$ : 2nd half.

| $\Theta_{i}=$ | $\pi\left(\Theta_{i}\right) \oplus c_{i}$ | $\pi\left(\Theta_{j 2}\right) \oplus c_{j 2}$ | $\pi\left(\Sigma_{j 2}\right) \oplus c_{j 2}$ | $\pi\left(\Sigma_{j 2}\right)$ | $\pi\left(\Theta_{j 2}\right)$ | $Y_{j 2, l}$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Cst $_{j 1}$ | $\Delta_{j 2}$ |  |  |  |  |  |
| $X_{j 1, l}$ | 19 | 20 | 21 | 22 | 23 | 24 |
| $\Theta_{j 1}$ | 27 | 28 | 29 | 30 | 26 | 25 |


[^0]:    ${ }^{3}$ We ignore the previous queries $X$ in the query computations and in the final output, as these are eventually defined recursively in terms of $Y$ and $m$.

[^1]:    ${ }^{4}$ This could be feasible as it is a collision probability for double-block construction. However, a term $\ell$ denoting the maximum message size may appear.

[^2]:    ${ }^{5}$ We implicitly fixed a primitive polynomial through which the multiplication is defined. In this paper, the whole analysis is independent of the choice of the polynomial and so we do not explicitly specify it.

[^3]:    ${ }^{6}$ Like collision relation, choice of $I$ is not unique. However, we implicitly fix a choice.
    ${ }^{7}$ For example, when $\mathcal{E}(x)=\left\{x_{i}=x_{j}: i \neq j\right\}$.

