

Combining Outputs of a Random Permutation: New Constructions and Tight Security Bounds by Fourier Analysis

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Abstract. We consider constructions that combine outputs of a single permutation $\pi : \{0, 1\}^n \rightarrow \{0, 1\}^n$ using a public function. These are popular constructions for achieving security beyond the birthday bound when implementing a pseudorandom function using a block cipher (i.e., a pseudorandom permutation). One of the best-known constructions (denoted SXoP[2, n]) XORs the outputs of 2 domain-separated calls to π . Modeling π as a uniformly chosen permutation, several previous works proved a tight information-theoretic indistinguishability bound for SXoP[2, n] of about $q/2^n$, where q is the number of queries. On the other hand, tight bounds are unknown for the generalized variant (denoted SXoP[r , n]) which XORs the outputs of $r > 2$ domain-separated calls to a uniform permutation.

In this paper, we obtain two results. Our first result improves the known bounds for SXoP[r , n] for all (constant) $r \geq 3$ (assuming $q \leq O(2^n/r)$ is not too large) in both the single-user and multi-user settings. In particular, for $q = 3$, our bound is about $\sqrt{u}q_{\max}/2^{2.5n}$ (where u is the number of users and q_{\max} is the maximal number of queries per user), improving the best-known previous result by a factor of at least 2^n .

For odd r , our bounds are tight for $q > 2^{n/2}$, as they match known attacks. For even r , we prove that our single-user bounds are tight by providing matching attacks.

Our second and main result is divided into two parts. First, we devise a family of constructions that output n bits by efficiently combining outputs of 2 calls to a permutation on $\{0, 1\}^n$, and achieve multi-user security of about $\sqrt{u}q_{\max}/2^{1.5n}$. Then, inspired by the CENC construction of Iwata [FSE'06], we further extend this family to output $2n$ bits by efficiently combining outputs of 3 calls to a permutation on $\{0, 1\}^n$. The extended construction has similar multi-user security of $\sqrt{u}q_{\max}/2^{1.5n}$. The new single-user ($u = 1$) bounds of $q/2^{1.5n}$ for both families should be contrasted with the previously best-known bounds of $q/2^n$, obtained by the comparable constructions of SXoP[2, n] and CENC.

All of our bounds are proved by Fourier analysis, extending the provable security toolkit in this domain in multiple ways.

1 Introduction

Efficient implementations of pseudorandom functions today typically use block ciphers, which are pseudorandom permutations that only achieve security up to

the birthday bound of $q = 2^{n/2}$ queries (where n is the block length). Since the security of many cryptosystems (such as encryption modes, MAC algorithms and authenticated encryption schemes) is based on pseudorandom functions, beyond-birthday bound security has become a popular research area, initiated in papers by Bellare, Krovetz, and Rogaway [2], and by Hall, Wagner, Kelsey, and Schneier [16].

1.1 XORing Permutation Outputs

One of the best-known constructions for achieving security beyond the birthday bound XORs the outputs of 2 permutations calls. This construction has two main variants. The first variant, denoted XoP[2, n] (XOR of Permutations), uses two permutations $\pi_1, \pi_2 : \{0, 1\}^n \mapsto \{0, 1\}^n$ to define $\text{XoP}[2, n]_{\pi_1, \pi_2} : \{0, 1\}^n \mapsto \{0, 1\}^n$ by $\text{XoP}[2, n]_{\pi_1, \pi_2}(i) = \pi_1(i) \oplus \pi_2(i)$. In practice, π_1 and π_2 are implemented using a block cipher, instantiated with independent keys. The second variant, denoted SXoP[2, n], uses 2 domain-separated calls to a single permutation $\pi : \{0, 1\}^n \mapsto \{0, 1\}^n$ to define $\text{SXoP}[2, n]_{\pi} : \{0, 1\}^{n-1} \mapsto \{0, 1\}^n$ by $\text{SXoP}[2, n]_{\pi}(i) = \pi(0\|i) \oplus \pi(1\|i)$ (where $\|$ denotes concatenation). As in the first variant, π is implemented using a block cipher. However, in information-theoretic security proofs, the block ciphers in both variants are replaced by idealized random permutations.

The second variant is more efficient in the sense that it only requires a single key. Yet, the advantage of the first variant is that it achieves better concrete security in idealized models.

Generalizations. Natural generalizations of the above variants XOR the outputs $r \geq 2$ permutations calls. The aim of these generalizations is to obtain even better security bounds.

In this paper, we are mainly interested in a generalization of the second variant, denoted SXoP[r , n]. It uses $r \geq 2$ domain-separated calls to a single permutation $\pi : \{0, 1\}^n \mapsto \{0, 1\}^n$ to define $\text{SXoP}[r, n]_{\pi} : \{0, 1\}^{n-\lceil \log r \rceil} \mapsto \{0, 1\}^n$ by $\text{SXoP}[r, n]_{\pi}(i) = \pi(0\|i) \oplus \pi(1\|i) \oplus \dots \oplus \pi(r-1\|i)$.

Previous results. Both variants have been analyzed in the idealized model by numerous papers in both the single-user and multi-user settings. The first variant (XoP) that uses independent permutations (and its generalized version) was analyzed in [7,6,8,9,10,12,20,21,23,25]. A tight security bound for XoP and its generalization was derived in [11] (also see [13] for XoP[2, n]), and further extended to the multi-user setting.

Works that analyzed the second variant SXoP (and its generalization) include [1,4,8,10,12,17,23,25]. In particular, for SXoP[2, n] a security bound of about $\frac{q}{2^n}$ was proved in [8,10,12]. This bound is tight as it is matched by a simple attack that checks whether the element 0 is output. The bound was extended to a tight bound in the multi-user setting (up to a logarithmic factor) in [17].

For the more general scheme SXoP $[r, n]$ with $r \geq 3$, tight bounds are unknown. The particular case of $r = 3$ was analyzed by Bhattacharya and Nandi in [4], deriving a bound of about $\frac{\sqrt{uq_{\max}}}{2^n}$ in the multi-user setting.

Remark 1. In practice, each permutation is instantiated with a keyed block cipher. In such computational settings, one needs to add an additional term (or terms) to the bounds derived above which take into account the optimal advantage in distinguishing the underlying block cipher (or block ciphers) from a uniformly chosen permutation (or permutations).

1.2 Iwata’s PRF construction

At FSE 2006 [18], Iwata introduced CENC, which is a beyond-birthday bound secure mode of operation, built from a PRF, $F[w, n] : \{0, 1\}^{n - \lceil \log(w+1) \rceil} \mapsto \{0, 1\}^{wn}$ using an underlying permutation $\pi : \{0, 1\}^n \mapsto \{0, 1\}^n$ and defined as

$$F[w, n]_{\pi}(i) = (\pi(0\|i) \oplus \pi(1\|i)) \parallel (\pi(0\|i) \oplus \pi(2\|i)) \parallel \dots \parallel (\pi(0\|i) \oplus \pi(w\|i)).$$

Thus, F only makes $w + 1$ calls to π in order to generate wn bits of output, whereas SXoP $[2, n]$ makes $2w$ calls to π .

When modeling π as an ideal permutation, [3,8,19] proved that $F[w, n]$ has an indistinguishability advantage upper bound of about $\frac{w^2 q}{2^n}$.

1.3 Our Results

In this paper, we obtain two results.

Result 1 - analysis of SXoP $[r, n]$. We improve known bounds for SXoP $[r, n]$ for all (constant) $r \geq 3$ (assuming $q \leq O(2^n/r)$ is not too large).

For odd r , we derive a bound of about $\frac{q}{2^{n(r-0.5)}}$ in the single-user setting and $\frac{\sqrt{uq}}{2^{n(r-0.5)}}$ in the multi-user setting. In particular, for $q = 3$, our bound $\frac{\sqrt{uq_{\max}}}{2^{2.5n}}$ improves the best-known previous one of [4] by a factor of at least 2^n . Our bounds for odd r are tight up to a constant factor (for $q \geq 2^{n/2}$), as they match attacks published by Patarin [24,26]. This includes the multi-user setting, where our bounds are matched by the simple generalization of the attacks of Patarin, which applies the single-user attack independently to each user and outputs a majority vote over the answers.

For even r , we prove a bound of about $\frac{q}{2^{nr/2}}$ in the single-user setting and an additional (slightly more complicated) bound of about $\min\left(\frac{\sqrt{uq_{\max}}}{2^{n(r/2-1/2)}}, \frac{uq_{\max}}{2^{nr/2}}\right)$ in the multi-user setting. Furthermore, we prove that our single-user bounds are tight by providing matching attacks, which improve the ones of [24,26].

Interestingly, our results show (for example) that SXoP $[3, n]$ (with a tight bound of $\frac{q}{2^{2.5n}}$) is provably more secure than SXoP $[4, n]$ (with a tight bound of $\frac{q}{2^{2n}}$). More generally, for odd $r \geq 3$, SXoP $[r, n]$ (with a bound of $\frac{q}{2^{n(r-0.5)}}$) is provably more secure than SXoP $[2r - 2, n]$ (with a bound of $\frac{q}{2^{n(r-1)}}$). Similar results hold in the multi-user setting.

Result 2 - definition and analysis of LXoP[L, n] and LXoP[$L, 2, n$].

LXoP[L, n]. We propose a family of constructions that output n bits by publicly combining outputs of 2 calls to a single permutation on $\{0, 1\}^n$, and achieve multi-user security of about $\frac{\sqrt{u}q_{\max}}{2^{1.5n}}$ (as long as $q_{\max} \leq O(2^n)$ is not too large). Hence, these constructions are provably secure up to $u = o(2^n)$ users for $q_{\max} \geq \Omega(2^n)$. Our (single-user) bound of $\frac{q}{2^n}$ improves upon the best previous bound of $\frac{q}{2^n}$ for a construction with similar parameters (obtained for SXoP[2, n]).

Our family of constructions is parameterized by a public linear orthomorphism, which is an invertible linear transformation $L : \{0, 1\}^n \rightarrow \{0, 1\}^n$ with the property that $L'(x) = x \oplus L(x)$ is itself a permutation. The construction is denoted by LXoP[L, n] and defined as LXoP[L, n] $_{\pi}(i) = \pi(0\|i) \oplus L(\pi(1\|i))$, where $i \in \{0, 1\}^{n-1}$.

It is easy to show that our bound $\frac{\sqrt{u}q_{\max}}{2^{1.5n}}$ is tight assuming $q \geq 2^{n/2}$ by similar attacks to the ones of [24,26]. Note that the bound we obtain is of the same order as the tight bound for XoP[2, n].

Importantly, there are many linear orthomorphisms $L : \{0, 1\}^n \rightarrow \{0, 1\}^n$ with the desired properties which are very simple and easy to implement. One example is $L(x^{(1)}, x^{(2)}) = (x^{(2)}, x^{(1)} \oplus x^{(2)})$, where $x^{(1)}, x^{(2)} \in \{0, 1\}^{n/2}$. Another example that may be more efficient to implement in hardware is $L(x) = (x \ggg 1) \oplus (x_1, 0, \dots, 0)$, i.e., cyclically rotate x by 1 bit to the right and XOR the first bit of x (denoted x_1) to the first bit of the result. Yet another example is doubling in the field \mathbb{F}_{2^n} . More details about linear orthomorphisms over \mathbb{F}_2^n can be found in [15].

Intuitively, the main reason that such constructions have a high security level is that (unlike SXoP[2, n]), every element generated by LXoP[L, n] is marginally uniform in $\{0, 1\}^n$. Indeed, let $x \in \{0, 1\}^n$ be such an element and write it as $x = y \oplus L(z)$, where $y, z \in \{0, 1\}^n$ are drawn uniformly without replacement. Then, fixing any $a \in \{0, 1\}^n$, the equality $x = a$ is equivalent to $y \oplus L(z) = a$. If $y, z \in \{0, 1\}^n$ were drawn uniformly and independently, then since L is invertible, the equation $y \oplus L(z) = a$ would have exactly 2^n solutions. However, since y, z are drawn uniformly without replacement, we subtract the solutions that satisfy $y = z$, and as L is an orthomorphism, the equation $y \oplus L(y) = a$ has exactly one solution. Consequently, for any $a \in \{0, 1\}^n$, the equation $y \oplus L(z) = a$ has exactly $2^n - 1$ solutions, namely, $x = y \oplus L(z)$ is uniformly distributed.

We remark that the use of linear orthomorphisms in cryptography (and particularly in the design of block ciphers) is not new. See [5] and references therein for examples. Hence, the main novelty of this work with respect to the LXoP[L, n] family (and its generalization below) is in the security proof, rather than the actual design.

LXoP[$L, 2, n$]. After analyzing LXoP[L, n], we extend the construction to obtain better efficiency by outputting $2n$ bits via 3 calls to the underlying permutation. Specifically, we define LXoP[$L, 2, n$] : $\{0, 1\}^{n-2} \mapsto \{0, 1\}^{2n}$ as

$$\text{LXoP}[L, 2, n]_{\pi}(i) = (\pi(0\|i) \oplus L(\pi(1\|i))) \parallel (\pi(1\|i) \oplus L(\pi(2\|i))).$$

We prove that $\text{LXoP}[L, 2, n]$ offers similar security to $\text{LXoP}[L, n]$ in both the single-user and multi-user settings, given that L is a linear orthomorphism. Compared to Iwata’s PRF [18], $\text{F}[2, n]$, the indistinguishability bound is improved from about $\frac{q}{2^n}$ to $\frac{q}{2^{3n/2}}$ (in the single-user setting), while having comparable parameters.

$\text{LXoP}[L, w, n]$. One can further extend LXoP to output wn bits via $w + 1$ permutation calls, similarly to Iwata’s PRF. Specifically, define

$$\text{LXoP}[L, w, n]_{\pi}(i) = (\pi(0\|i) \oplus L(\pi(1\|i))) \parallel \dots \parallel (\pi(w - 1\|i) \oplus L(\pi(w\|i))),$$

where $i \in \{0, 1\}^{n - \lceil \log(w+1) \rceil}$. To achieve high security, we require that the iterated invertible linear function L^j has no short cycles of length up to w , namely for every $x \in \{0, 1\}^n$ such that $x \neq 0$ and $1 \leq j \leq w$, $x \oplus L^j(x) \neq 0$. Such efficient functions L are easy to build (e.g., from linear-feedback shift registers).

While it is not difficult (albeit somewhat technical) to extend our security analysis of $\text{LXoP}[L, 2, n]$ to $\text{LXoP}[L, w, n]$ for very small values of $w > 2$, the analysis for general w is more involved and we leave it to future work.

We remark that a different variant of $\text{LXoP}[L, w, n]_{\pi}(i)$ defines the j ’s output block (for $j = 1, \dots, w$) as $L^j(\pi(0\|i)) \oplus \pi(j\|i)$. However, this variant seems to be inferior to the one above in terms of both security (for large w) and efficiency, since the computations of $L^j(\pi(0\|i))$ for different values of j are more difficult to parallelize.

1.4 Technical Overview

Similarly to the previous works [11,13,14], we prove our results by Fourier analysis. We start by elaborating on the techniques of [11,13] that are relevant to this paper.

Previous techniques [11,13]. First, the distinguishing advantage of the adversary is bounded by the statistical distance between the distribution generated by the analyzed construction and the uniform distribution. Consider a sample from a distribution generated by the analyzed construction, which is over $\mathbb{F}_2^{q \times n}$ (i.e., composed of q elements in $\{0, 1\}^n$). The statistical distance of this distribution from the uniform distribution can be bounded in the “Fourier domain” by bounding the bias (i.e., Fourier coefficient) of each of the 2^{qn} possible masks (i.e., linear equations over \mathbb{F}_2) applied to the bits of the sample.

In [11,13], the task of bounding the Fourier coefficients for the distribution function generated by the XoP construction was reduced to the task of bounding the Fourier coefficients for the distribution generated by the underlying primitive, namely, a random permutation. This reduction was based on the fact that XORing together samples generated by independent random permutations corresponds to a convolution operation, which is simple multiplication in the Fourier domain.

Considering k elements (for any $1 \leq k \leq q$) drawn uniformly without replacement, the proof of [11] used bounds on two quantities of Fourier coefficients on masks that involve all of these k elements (called level- k coefficients).

1. The maximal level- k Fourier coefficient in absolute value.
2. The level- k Fourier weight, which is equal to the sum of squares of all level- k Fourier coefficients.

Our techniques. We would like to use a similar approach to bound the distinguishing advantage of the adversary against the SXoP and LxoP constructions. However, unlike the XoP construction, these do not involve XORing together independent permutations. Therefore, the step that reduces the analysis to bounding the Fourier coefficients of a random permutation via convolution is not applicable anymore.

Nevertheless, we prove that the Fourier coefficients of the distribution generated by the SXoP and LxoP constructions are, in fact, structured subsets of the Fourier coefficients of a random permutation.

For example, denote by $x \in \{0, 1\}^n$ a single element of a sample generated by SXoP[2, n]. Consider a mask involving a single element $\alpha \in \{0, 1\}^n \neq 0$ (i.e., a mask of level 1), and assume we wish to analyze the bias of the linear equation $\alpha_1 x_1 \oplus \dots \oplus \alpha_n x_n$. Since x is generated by SXoP[2, n], we can write $x = y \oplus z$, where $y, z \in \{0, 1\}^n$ are generated by a random permutation. The above linear equation can therefore be written as $\alpha_1 (y_1 \oplus z_1) \oplus \dots \oplus \alpha_n (y_n \oplus z_n) = (\alpha_1 y_1 \oplus \dots \oplus \alpha_n y_n) \oplus (\alpha_1 z_1 \oplus \dots \oplus \alpha_n z_n)$. The bias of this equation is exactly the Fourier coefficient of a random permutation on the level-2 symmetric mask $(\alpha, \alpha) \in \{0, 1\}^{2n}$.

In general, level- k Fourier coefficients of the distribution generated by SXoP[r, n] correspond to symmetric level- (rk) Fourier coefficients of a random permutation. One can similarly prove that level- k Fourier coefficients of the distribution generated by LxoP[L, n] correspond to level- $2k$ Fourier coefficients of a random permutation (with a certain structure that depends on L). A similar property also holds for LxoP[$L, 2, n$]. Therefore, we can use the two bounds above on the Fourier coefficients of a random permutation to analyze the distributions generated by the SXoP and LxoP constructions.

Framework for bounding Fourier weight of sampling without replacement on structured subsets of masks. Unfortunately, using the general level- k bounds naively is not sufficient to obtain tight indistinguishability bounds for the constructions we analyze, particularly for LxoP. Essentially, the general level- k bound on the weight (i.e., the second bound) is tight for dense subsets of masks that contain (a large fraction of) all level- k masks. However, the subsets we need to analyze are structured and very sparse.

As a result, in this paper we develop a framework that allows to bound the Fourier weight of the sampling without replacement density function (normalized distribution function) on structured subsets of masks. The framework takes into

account the particular structure of the subset and significantly improves the naive bounds for the constructions we analyze.

Technically, the framework uses a (known) recursive formula for calculating the Fourier coefficient on any single mask α as a sum of Fourier coefficients on lower-level masks, derived from α . We show how to manipulate the formula to collectively analyze the Fourier weight of a subset of masks that have a common structure, determined by the construction we analyze. Specifically, each recursive call bounds the weight of an increasingly denser subset of masks, and we apply the general bounds only at the leaves of the recursion tree, where they are closer to being tight. The power and generality of this framework is demonstrated by applying it to obtain tight indistinguishability bounds for all constructions we analyze in this paper.

A notable exception to the above is the SXoP $[r, n]$ construction with even r , whose analysis requires an additional central technical contribution, summarized below.

Mixed \mathcal{L}^1 and \mathcal{L}^2 bounds. For the SXoP $[r, n]$ construction with even r the above strategy is not sufficient to obtain tight indistinguishability bounds. Essentially, this is because of a quadratic loss of the standard Cauchy-Schwarz inequality that bounds the statistical distance (\mathcal{L}^1 distance) of the analyzed distribution to the uniform distribution using the \mathcal{L}^2 distance. In order to overcome this loss, we bound the statistical distance by a mixture of \mathcal{L}^1 and \mathcal{L}^2 bounds using the Fourier decomposition of the distribution (density) function. While such mixed bounds have been used before in a hybrid argument (e.g., in [10]), we stress that our mixed bounds are purely analytical in the sense that the “hybrids” that we use do not necessarily correspond to actual distributions, but rather to a Fourier decomposition of the density function.

An additional advantage of this technique is that it allows to *lower bound* the statistical distance, i.e., analyze the optimal attack in the Fourier domain using the reverse triangle inequality. Indeed, the optimal attack against the SXoP $[r, n]$ construction reveals itself during the analysis of the level-1 Fourier coefficients, and it simply corresponds to comparing the number of 0 elements in the sample to a threshold.

1.5 Paper Structure

The rest of this paper is organized as follows. Next, in Section 2, we describe preliminaries. In Section 4 we prove our results regarding the SXoP constructions, while in Section 5 and Section 6 we analyze the variants of the LXoP construction.

2 Preliminaries

For a positive integer m (i.e., $m \in \mathbb{Z}^{\geq 1}$), denote $[m] = \{1, 2, \dots, m\}$. For $m_1, m_2 \in \mathbb{Z}$ such that $m_1 \leq m_2$, denote $[m_1, m_2] = \{m_1, m_1 + 1, \dots, m_2\}$.

For a set \mathcal{A} , denote its size by $|\mathcal{A}|$. For any integer $k > 0$ and a real number t , define the falling factorial as $(t)_k = t(t-1)\dots(t-(k-1))$. Further define $(t)_0 = 1$.

Let $n, m \in \mathbb{Z}^{\geq 1}$ such that $n \geq m$. Then, $(\frac{n}{m})^m \leq \binom{n}{m} \leq (\frac{e \cdot n}{m})^m$.

Proposition 1. *Let $a, b, c, d, k \in \mathbb{R}^{\geq 0}$. Define the functions $B(k) = (ak+b)^{ck+d}$ and $C(k) = \frac{1}{(b-ak)^{ck+d}}$. Then*

$$\frac{B(k+1)}{B(k)} \leq (a(k+1)+b)^c e^{\frac{a(ck+d)}{ak+b}}, \text{ and } \frac{C(k+1)}{C(k)} \leq \frac{1}{(b-a(k+1))^c} e^{\frac{a(ck+d)}{b-a(k+1)}},$$

where the last inequality assumes $b > a(k+1)$.

Proof. We have

$$\begin{aligned} \frac{B(k+1)}{B(k)} &= \frac{(a(k+1)+b)^{c(k+1)+d}}{(ak+b)^{ck+d}} = \frac{(a(k+1)+b)^{c(k+1)+d}}{(a(k+1)+b)^{ck+d}} \frac{(a(k+1)+b)^{ck+d}}{(ak+b)^{ck+d}} \\ &= (a(k+1)+b)^c \left(1 + \frac{a}{ak+b}\right)^{ck+d} \leq (a(k+1)+b)^c e^{\frac{a(ck+d)}{ak+b}}. \end{aligned}$$

and

$$\begin{aligned} \frac{C(k+1)}{C(k)} &= \frac{(b-a(k+1))^{-c(k+1)-d}}{(b-ak)^{-ck-d}} = \frac{(b-a(k+1))^{-c(k+1)-d}}{(b-a(k+1))^{-ck-d}} \frac{(b-a(k+1))^{-ck-d}}{(b-ak)^{-ck-d}} \\ &= \frac{1}{(b-a(k+1))^c} \left(1 + \frac{a}{b-a(k+1)}\right)^{ck+d} \leq \frac{1}{(b-a(k+1))^c} e^{\frac{a(ck+d)}{b-a(k+1)}}. \end{aligned}$$

■

Let x be an element (from an arbitrary domain) and let $m \in \mathbb{Z}^{\geq 1}$. Define $x^{\circ m} = \underbrace{(x, \dots, x)}_m$ to be the sequence of m repetitions of x . For a sequence

(x_1, \dots, x_k) , define $(x_1, \dots, x_k)^{\circ m} = ((x_1)^{\circ m}, \dots, (x_k)^{\circ m})$.

Let $m \in \mathbb{Z}^{\geq 1}$. We denote the sequence of elements (x_1, \dots, x_m) by $x_{1..m}$. Similarly, the sequence of elements (x^1, \dots, x^m) is denoted by $x^{1..m}$. Furthermore, for $m_1, m_2 \in \mathbb{Z}^{\geq 1}$, denote the sequence of $m_1 m_2$ elements $(x_1^1, \dots, x_1^{m_2}, \dots, x_{m_1}^1, \dots, x_{m_1}^{m_2})$ by $x_{1..m_1}^{1..m_2}$.

Let \mathbb{F} be a field and $v \in \mathbb{F}^{k_1 \times k_2}$ a matrix of elements in \mathbb{F} . We index the elements of v in a natural way, namely, for $i \in [k_1]$, $v_i \in \mathbb{F}^{k_2}$ is the i 'th row of v and for $j \in [k_2]$, $v_{i,j} \in \mathbb{F}$ is its j 'th entry.

For two (row) vectors $v, u \in \mathbb{F}^k$, we denote by $\langle u, v \rangle_{\mathbb{F}} = u \cdot v^T = \sum_{i \in [k]} u_i v_i$ their inner product (where v^T is the transpose of v and addition and multiplication are over \mathbb{F}). Similarly, for matrices $v, u \in \mathbb{F}^{k_1 \times k_2}$, define $\langle u, v \rangle_{\mathbb{F}} = \sum_{i \in [k_1]} u_i \cdot (v_i)^T = \sum_{(i,j) \in [k_1] \times [k_2]} u_{i,j} v_{i,j}$.

In this paper, we typically deal with matrices $x \in \mathbb{F}_2^{k \times n}$, where n is considered a parameter and k may vary. We denote $N = 2^n$.

Let $L \in \mathbb{F}_2^{n \times n}$. Denote by L^T the transpose of L . Further, let $x \in \mathbb{F}_2^{k \times n}$. We define $L(x) \in \mathbb{F}_2^{k \times n}$ by $L(x)_i = x_i \cdot L$ for $i \in [k]$ (where we view x_i as a row vector in \mathbb{F}_2^n , multiplied with L).

Asymptotic notation. While all of our results are fully explicit, we sometimes use standard asymptotic notation to give intuition about the bounds we obtain. In particular, we use the notation $O_r(\cdot)$ and $\Omega_r(\cdot)$ that suppress arbitrary functions of r (as we mostly think of it as a small constant).

2.1 Probability

Definition 1 (Density function). A (probability) density function on $\mathbb{F}_2^{q \times n}$ is a nonnegative function $\varphi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ satisfying $\mathbb{E}_{x \in \mathbb{F}_2^{q \times n}}[\varphi(x)] = 1$, where $x \in \mathbb{F}_2^{q \times n}$ is uniformly chosen.

We write $x \sim \varphi$ to denote that x is a sample drawn from the associated probability distribution, defined by $\Pr_{x \sim \varphi}[x = y] = \frac{\varphi(y)}{2^{qn}}$ for every $y \in \mathbb{F}_2^{q \times n}$. In particular, the uniform probability density function over $\mathbb{F}_2^{q \times n}$ is the constant function 1, and we denote it by $\mathbf{1}_{qn}$.

Let $\mathcal{A} \subseteq \mathbb{F}_2^{q \times n}$. We write $x \sim \mathcal{A}$ to denote that x is selected uniformly at random from \mathcal{A} .

Proposition 2 ([22], Fact 1.21). If $\varphi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ is a density function and $f : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}$, then $\mathbb{E}_{x \sim \varphi}[f(x)] = \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}}[\varphi(x)f(x)]$.

Definition 2 (Statistical distance). The statistical distance between two probability density functions $\varphi, \psi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ is

$$\text{SD}(\varphi, \psi) = \frac{1}{2} \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} |\varphi(x) - \psi(x)|.$$

2.2 Fourier Analysis

We define the Fourier-Walsh expansion of functions on the Boolean cube, adapted to our setting, and state the basic results that we will use. These results are mostly taken from [22].

Definition 3 (Fourier expansion). Given $\alpha \in \mathbb{F}_2^{q \times n}$, define $\chi_\alpha : \mathbb{F}_2^{q \times n} \mapsto \{-1, 1\}$ by

$$\chi_\alpha(x) = (-1)^{\langle \alpha, x \rangle_{\mathbb{F}_2}} = \prod_{i \in [q]} (-1)^{\langle \alpha_i, x_i \rangle_{\mathbb{F}_2}} = \prod_{i \in [q], j \in [n]} (-1)^{\alpha_{i,j} \cdot x_{i,j}}.$$

The set $\{\chi_\alpha\}_{\alpha \in \mathbb{F}_2^{q \times n}}$ is an orthonormal basis for the set of functions $\{f \mid f : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}\}$, with respect to the normalized inner product $\frac{1}{|\mathbb{F}_2^{q \times n}|} \langle f, g \rangle_{\mathbb{R}} = \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}}[f(x)g(x)]$. Hence each $\{f \mid f : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}\}$ can be decomposed to

$$f = \sum_{\alpha \in \mathbb{F}_2^{q \times n}} \widehat{f}(\alpha) \chi_\alpha,$$

where $\widehat{f}(\alpha) = \mathbb{E}[\chi_\alpha f]$, and in particular, $\widehat{f}(0) = \mathbb{E}[f]$.

Each element in $\{\chi_\alpha\}_{\alpha \in \widehat{\mathbb{F}}_2^{q \times n}}$ is called a *character*. We refer to α as a *mask*, and to $\widehat{f}(\alpha)$ as the *Fourier coefficient of f on α* . To distinguish the domain of characters from the input domain we write it as $\widehat{\mathbb{F}}_2^{q \times n}$, and thus $f(x) = \sum_{\alpha \in \widehat{\mathbb{F}}_2^{q \times n}} \widehat{f}(\alpha) \chi_\alpha(x)$. For a mask $\alpha \in \widehat{\mathbb{F}}_2^{q \times n}$, we write

$$\text{supp}(\alpha) = \{i \mid \alpha_i \neq 0\} \text{ and } \#\alpha = |\text{supp}(\alpha)|.$$

We call $\#\alpha$ the *level* of α , and $\widehat{f}(\alpha)$ is a Fourier coefficient of level $\#\alpha$. Throughout the paper, we mostly use the shorter notation $\mathcal{NZ}_\alpha = \text{supp}(\alpha)$.

For integer parameters $n \geq 1$ and $0 \leq k_0 \leq k_1$, we define the sets of masks $\mathcal{M}_{=k_0, k_1}^n = \{\alpha \in \widehat{\mathbb{F}}_2^{k_1 \times n} \mid \#\alpha = k_0\}$, and $\mathcal{M}_{\geq k_0, k_1}^n = \{\alpha \in \widehat{\mathbb{F}}_2^{k_1 \times n} \mid \#\alpha \geq k_0\}$.

Definition 4 (Fourier weight and maximal magnitude). For a function $f : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}$, we define the Fourier weight of f at level k to be

$$\mathbb{W}^{=k}[f] = \sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{q \times n} \\ \#\alpha = k}} \widehat{f}(\alpha)^2 = \sum_{\alpha \in \mathcal{M}_{=k, q}^n} \widehat{f}(\alpha)^2.$$

The maximal magnitude of a level- k Fourier coefficient of f is

$$\mathbb{M}^{=k}[f] = \max_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{q \times n} \\ \#\alpha = k}} \{|\widehat{f}(\alpha)|\} = \max_{\alpha \in \mathcal{M}_{=k, q}^n} \{|\widehat{f}(\alpha)|\}.$$

Proposition 3 ([22], Proposition 1.13 – variance). The variance of $f : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}$ is

$$\text{Var}[f] = \mathbb{E}[f^2] - \mathbb{E}[f]^2 = \sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{q \times n} \\ \alpha \neq 0}} \widehat{f}(\alpha)^2 = \sum_{k=1}^q \mathbb{W}^{=k}[f].$$

Proposition 4 ([22], Exercise 1.23 – bound on statistical distance from uniform by variance). Let $\varphi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ be a density function. Then

$$\text{SD}(\varphi, \mathbf{1}_{qn}) \leq \frac{1}{2} \sqrt{\text{Var}[\varphi]}.$$

We generalize this bound below to a combination of \mathcal{L}^1 and \mathcal{L}^2 distances in the Fourier domain.

Proposition 5 (Bidirectional bounds on statistical distance from uniform by \mathcal{L}^1 and \mathcal{L}^2 distances). Let $\varphi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ be a density function. Let $\mathcal{S} \subset \widehat{\mathbb{F}}_2^{q \times n}$ be any set of masks, which does not contain the zero mask. Let $\overline{\mathcal{S}} = \widehat{\mathbb{F}}_2^{q \times n} \setminus \{\mathcal{S} \cup \{0\}\}$ be the complementary set of masks (not including the zero mask). Then

$$-\sqrt{\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha)^2} \leq 2 \text{SD}(\varphi, \mathbf{1}_{qn}) - \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \leq \sqrt{\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha)^2}.$$

Note that setting $\mathcal{S} = \emptyset$ above gives Proposition 4, hence it is indeed a generalization (the lower bound on $\text{SD}(\varphi, \mathbf{1}_{qn})$ in this case is trivial). In general, Proposition 5 can give better results than Proposition 4 in case $\sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x)$ has a significant amount of cancellations due to opposite signs of the terms $\widehat{\varphi}(\alpha) \chi_\alpha(x)$ (on average over $x \sim \mathbb{F}_2^{q \times n}$).

Proof. We have

$$\begin{aligned} 2\text{SD}(\varphi, \mathbf{1}_{qn}) &= \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} |\varphi(x) - 1| = \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\substack{\alpha \in \mathbb{F}_2^{q \times n} \\ \alpha \neq 0}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \\ &= \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) + \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \\ &\leq \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| + \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \end{aligned}$$

For the upper bound, it remains to prove that $\mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \leq \sqrt{\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha)^2}$. Applying the Cauchy-Schwarz inequality,

$$\begin{aligned} \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| &\leq \sqrt{\mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left[\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right]^2} \\ &= \sqrt{\sum_{(\alpha, \beta) \in \overline{\mathcal{S}} \times \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \widehat{\varphi}(\beta) \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} [\chi_\alpha(x) \chi_\beta(x)]} = \sqrt{\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha)^2}, \end{aligned}$$

where the final equality is by orthogonality of the characters.

For the lower bound, observe similarly that

$$\begin{aligned} 2\text{SD}(\varphi, \mathbf{1}_{qn}) &= \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) + \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \\ &\geq \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| - \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| \\ &\geq \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{S}} \widehat{\varphi}(\alpha) \chi_\alpha(x) \right| - \sqrt{\sum_{\alpha \in \overline{\mathcal{S}}} \widehat{\varphi}(\alpha)^2}. \end{aligned}$$

■

We state an additional basic result regarding variance.

Proposition 6 ([11], Proposition 6 – Variance of independent samples). *Let $\varphi : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ be a density function. Let $u \geq 1$ be an integer and let $\varphi^{\times u} : \mathbb{F}_2^{(qu) \times n} \mapsto \mathbb{R}^{\geq 0}$ be the density function obtained by concatenating u independent samples drawn from φ . Then,*

$$\text{Var}[\varphi^{\times u}] \leq 2u \text{Var}[\varphi], \text{ assuming } u \text{Var}[\varphi] \leq \frac{1}{2}.$$

2.3 Cryptographic Preliminaries

We use the standard notion of PRF security, as defined below. Let $H : \mathcal{K} \times \{0, 1\}^{m_1} \mapsto \{0, 1\}^{m_2}$ be a family of functions and $\text{Func}(m_1, m_2)$ be the set of all functions $g : \{0, 1\}^{m_1} \mapsto \{0, 1\}^{m_2}$. Let A be an algorithm with oracle access to a function $f : \{0, 1\}^{m_1} \mapsto \{0, 1\}^{m_2}$. The PRF advantage of A against H is

$$\text{Adv}_H^{\text{prf}}(A) = \left| \Pr_{K \sim \mathcal{K}} [A^{H_K(\cdot)} \Rightarrow 1] - \Pr_{f \sim \text{Func}(m_1, m_2)} [A^{f(\cdot)} \Rightarrow 1] \right|.$$

We further define the optimal advantage

$$\text{Opt}_H^{\text{prf}}(q) = \max\{\text{Adv}_H^{\text{prf}}(A) \mid A \text{ makes } q \text{ queries}\}.$$

In the multi-user setting we have u users, each with an independent instantiation of the cryptosystem. The adversary can issue (up to) q_{\max} queries to each user with the goal of distinguishing the u instantiations of the cryptosystem from u instantiations of a random function. We define the PRF advantage of A against H in the multi-user setting as

$$\begin{aligned} \text{Adv}_{H,u}^{\text{mu-prf}}(A) = & \left| \Pr_{K_1, \dots, K_u \sim \mathcal{K}} [A^{H_{K_1}(\cdot), \dots, H_{K_u}(\cdot)} \Rightarrow 1] \right. \\ & \left. - \Pr_{f_1, \dots, f_u \sim \text{Func}(m_1, m_2)} [A^{f_1(\cdot), \dots, f_u(\cdot)} \Rightarrow 1] \right| \end{aligned}$$

We further define the optimal advantage

$$\text{Opt}_{H,u}^{\text{mu-prf}}(q_{\max}) = \max\{\text{Adv}_{H,u}^{\text{mu-prf}}(A) \mid A \text{ makes } q_{\max} \text{ queries to each user}\}.$$

Bounding the optimal advantage using Fourier analysis. In this paper we will consider families of functions of the form $H : \mathcal{K} \times \{0, 1\}^m \mapsto \{0, 1\}^n$ with the property that the output distribution is independent of the queries of the adversary. Thus, we ignore these queries and focus on analyzing the output distribution (density function) generated by H . Given that the adversary makes q queries to H , we may denote the density function generated by H as $\varphi_{H(n,q)} : \mathbb{F}_2^{q \times n} \rightarrow \mathbb{R}^{\geq 0}$.

By well-known properties of the statistical distance, the advantage of the optimal distinguisher against H is equal to the statistical distance of φ_H from uniform, namely,

$$\text{Opt}_H^{\text{prf}}(q) = \text{SD}(\varphi_{H(n,q)}, \mathbf{1}_{qn}). \quad (1)$$

In the multi-user setting, an adversary against H obtains a sample of $(\varphi_{H(n,q)})^{\times u} : \mathbb{F}_2^{(q_{\max} u) \times n} \mapsto \mathbb{R}^{\geq 0}$, where $(\varphi_{H(n,q)})^{\times u}$ is the density function obtained by concatenating u independent samples drawn from $\varphi_{H(n,q)}$. Similarly to the single-user setting,

$$\text{Opt}_{H,u}^{\text{mu-prf}}(q_{\max}) = \text{SD}((\varphi_{H(n,q_{\max})})^{\times u}, \mathbf{1}_{uq_{\max}n}). \quad (2)$$

In this paper, we mostly bound the optimal advantage by bounding $\text{Var}[\varphi_{H(n,q)}]$ using the following basic result.

Proposition 7 (Bounds on advantage using variance). *Assume that the output distribution generated by $H : \mathcal{K} \times \{0, 1\}^m \mapsto \{0, 1\}^n$ is independent of the queries of the adversary. Denote by $\varphi_{H(n,q)} : \mathbb{F}_2^{q \times n} \rightarrow \mathbb{R}^{\geq 0}$ the density function generated by H . Then,*

$$\text{Opt}_H^{\text{prf}}(q) \leq \frac{1}{2} \sqrt{\text{Var}[\varphi_{H(n,q)}]}, \text{ and } \text{Opt}_{H,u}^{\text{mu-prf}}(q_{\max}) \leq \frac{1}{\sqrt{2}} \sqrt{u \text{Var}[\varphi_{H(n,q_{\max})}]},$$

assuming $u \text{Var}[\varphi_{H(n,q_{\max})}] \leq \frac{1}{2}$, or equivalently, $\frac{1}{\sqrt{2}} \sqrt{u \text{Var}[\varphi_{H(n,q_{\max})}]} \leq \frac{1}{2}$.

Proof. First, by (1) and Proposition 4,

$$\text{Opt}_H^{\text{prf}}(q) = \text{SD}(\varphi_{H(n,q)}, \mathbf{1}_{qn}) \leq \frac{1}{2} \sqrt{\text{Var}[\varphi_{H(n,q)}]}.$$

Second, by (2), Proposition 4 and Proposition 6,

$$\begin{aligned} \text{Opt}_{H,u}^{\text{mu-prf}}(q_{\max}) &= \text{SD}((\varphi_{H(n,q_{\max})})^{\times u}, \mathbf{1}_{uq_{\max}n}) \\ &\leq \frac{1}{2} \sqrt{\text{Var}[\varphi_{H(n,q_{\max})}^{\times u}]} \leq \frac{1}{\sqrt{2}} \sqrt{u \text{Var}[\varphi_{H(n,q_{\max})}]}. \end{aligned}$$

■

Symmetric properties. In addition to the output distribution being independent of the queries of the adversary, all the functions $H : \mathcal{K} \times \{0, 1\}^m \mapsto \{0, 1\}^n$ we analyze in this paper are symmetric in the following sense: if $x \sim \varphi_{H(n,q)}$, then for every set of k distinct indices $\{i_1, i_2, \dots, i_k\} \subseteq [q]$, $(x_{i_1}, \dots, x_{i_k})$ are k elements that are marginally sampled from $\varphi_{H(n,k)}$, namely, $(x_{i_1}, \dots, x_{i_k}) \sim \varphi_{H(n,k)}$. Therefore, for $1 \leq k \leq q$, we have $\text{M}^k[\varphi_{H(n,q)}] = \text{M}^k[\varphi_{H(n,k)}]$ and

$$\begin{aligned} \text{W}^k[\varphi_{H(n,q)}] &= \sum_{\alpha \in \mathcal{M}_{k,q}^n} \widehat{\varphi}_{H(n,q)}(\alpha)^2 = \sum_{\{i_1, \dots, i_k\} \subseteq [q] \text{ distinct}} \sum_{\substack{\beta \in \widehat{\mathbb{F}}_2^{k \times n} \\ \text{supp}(\beta) = \{i_1, \dots, i_k\}}} \widehat{\varphi}_{H(n,k)}(\beta)^2 \\ &= \sum_{\{i_1, \dots, i_k\} \subseteq [q] \text{ distinct}} \text{W}^k[\varphi_{H(n,k)}] = \binom{q}{k} \text{W}^k[\varphi_{H(n,q)}]. \end{aligned}$$

These symmetric properties are repeatedly used throughout the paper (often without explicitly referring to them). Another result on symmetric functions (which we do not explicitly use) is given in Appendix A.

Sampling without replacement. We define the density function of sampling without replacement.

Definition 5 (Density function of sampling without replacement). *For positive integers n, q such that $1 \leq q \leq 2^n$, let $\mu_{n,q} : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ be the density function associated with the process of uniformly sampling q elements from \mathbb{F}_2^n without replacement. Specifically, for $x \in \mathbb{F}_2^{q \times n}$,*

$$\mu_{n,q}(x) = \begin{cases} \frac{N^q}{(N)_q} & \text{if } x_i \neq x_j \text{ for all } i, j \in [q] \text{ (} i \neq j \text{)}, \\ 0 & \text{otherwise.} \end{cases}$$

Furthermore, define $\mu_{n,0}$ to be the constant 1.

The SXoP[r, n] construction. Let $\text{Perm}(n)$ be the set of all permutations on $\{0, 1\}^n$ (i.e., the set of all $\pi : \{0, 1\}^n \mapsto \{0, 1\}^n$). For positive integers r, n such that $r \geq 2$, define the family of functions $\text{SXoP}[r, n] : (\text{Perm}(n)) \times \{0, 1\}^{n - \lceil \log r \rceil} \mapsto \{0, 1\}^n$ by

$$\text{SXoP}[r, n]_{\pi}(i) = \pi(0\|i) \oplus \pi(1\|i) \oplus \dots \oplus \pi(r-1\|i),$$

where in $\pi(j\|i)$, $j \in \{0, 1\}^{\lceil \log r \rceil}$ is encoded in binary for $j = 0, \dots, r-1$, and $\|$ denotes concatenation. We will be interested in bounding $\text{Opt}_{\text{SXoP}[r, n]}^{\text{prf}}(q)$ as a function of the parameters r, n, q (and deriving similar bounds in the multi-user setting). By symmetry of the randomly chosen permutation π , an adversary against $\text{SXoP}[r, n]$ obtains the XOR of r samples, each containing q elements of $\{0, 1\}^n$, where all rq elements are chosen uniformly without replacement (regardless of the actual queries).

Let $\nu_{n, q}^{(r)} : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ denote the density function of a sample generated by the $\text{SXoP}[r, n]$ construction.

The LXoP[L, n] and LXoP[$L, 2, n$] constructions. Let $L \in \mathbb{F}_2^{m \times n}$ be an invertible matrix. Define the family of functions $\text{LXoP}[L, n] : (\text{Perm}(n)) \times \{0, 1\}^{n-1} \mapsto \{0, 1\}^n$ by

$$\text{LXoP}[L, n]_{\pi}(i) = \pi(0\|i) \oplus L(\pi(1\|i)).$$

Moreover, define the family of functions $\text{LXoP}[L, 2, n] : (\text{Perm}(n)) \times \{0, 1\}^{n-2} \mapsto \{0, 1\}^{2n}$ by

$$\text{LXoP}[L, 2, n]_{\pi}(i) = (\pi(0\|i) \oplus L(\pi(1\|i))) \parallel (\pi(1\|i) \oplus L(\pi(2\|i))).$$

We will be interested in bounding $\text{Opt}_{\text{LXoP}[L, n]}^{\text{prf}}(q)$ and $\text{Opt}_{\text{LXoP}[L, 2, n]}^{\text{prf}}(q)$ as a function of the parameters n, q (and deriving similar bounds in the multi-user setting). As in the case of $\text{SXoP}[r, n]$, the distributions generated by $\text{LXoP}[L, n]$ and $\text{LXoP}[L, 2, n]$ are independent of the queries of the adversary. Let $\xi_{n, q}^{(L)} : \mathbb{F}_2^{q \times n} \mapsto \mathbb{R}^{\geq 0}$ and $\xi_{n, 2, q}^{(L)} : \mathbb{F}_2^{q \times 2n} \mapsto \mathbb{R}^{\geq 0}$ denote the density functions of samples generated by the $\text{LXoP}[L, n]$ and $\text{LXoP}[L, 2, n]$ constructions, respectively.

2.4 Fourier Properties of $\mu_{n, k}$

We use several results about Fourier properties of $\mu_{n, k}$, mostly taken from [11, 13].

Proposition 8 ([11], Proposition 12 – Permuting elements preserves Fourier coefficients). *Let $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$. Let $\pi : [k] \mapsto [k]$ be a permutation and define the mask $\alpha^{\pi} \in \widehat{\mathbb{F}}_2^{k \times n}$ by $(\alpha^{\pi})_i = \alpha_{\pi(i)}$ for $i \in [k]$. Then, $\widehat{\mu}_{n, k}(\alpha^{\pi}) = \widehat{\mu}_{n, k}(\alpha)$.*

Proposition 8 is repeatedly used throughout the paper (often without explicitly referring to it).

Proposition 9. For any $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$ such that $\bigoplus_{i \in [k]} \alpha_i \neq 0$ we have $\widehat{\mu}_{n,k}(\alpha) = 0$.

Proof. Let $y \in \mathbb{F}_2^n$ be arbitrary. Observe that for $x \in \mathbb{F}_2^{k \times n}$, $\mu_{n,k}(x) = \mu_{n,k}(x_1 \oplus y, \dots, x_k \oplus y)$. Therefore,

$$\begin{aligned} \widehat{\mu}_{n,k}(\alpha) &= \mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} [\mu_{n,k}(x) \chi_\alpha(x)] = \mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} [\mu_{n,k}(x_1 \oplus y, \dots, x_k \oplus y) \chi_\alpha(x_1, \dots, x_k)] \\ &= \mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} [\mu_{n,k}(x_1 \oplus y, \dots, x_k \oplus y) \chi_\alpha(x_1 \oplus y, \dots, x_k \oplus y)] \chi_{(\bigoplus_{i \in [k]} \alpha_i)}(y) \\ &= \mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} [\mu_{n,k}(x) \chi_\alpha(x)] \chi_{(\bigoplus_{i \in [k]} \alpha_i)}(y) = \widehat{\mu}_{n,k}(\alpha) \chi_{(\bigoplus_{i \in [k]} \alpha_i)}(y). \end{aligned}$$

If $\widehat{\mu}_{n,k}(\alpha) \neq 0$, we divide both sides by $\widehat{\mu}_{n,k}(\alpha)$. We deduce that for every $y \in \mathbb{F}_2^n$, $\chi_{(\bigoplus_{i \in [k]} \alpha_i)}(y) = 1$, implying that $\bigoplus_{i \in [k]} \alpha_i = 0$. \blacksquare

The following is a recursive formula for $\mu_{n,k}(\alpha)$.

Proposition 10 ([13], Section 4 – recursive formula for $\mu_{n,k}(\alpha)$). For parameters $k_1 \geq k_0 \geq 2$, let $\alpha \in \widehat{\mathbb{F}}_2^{k_1 \times n}$ with $\#\alpha = k_0$, and define $\mathcal{NZ}_\alpha = \{i \in [k_1] \mid \alpha_i \neq 0\}$ ($|\mathcal{NZ}_\alpha| = k_0$). Then for any $j \in \mathcal{NZ}_\alpha$,

$$\widehat{\mu}_{n,k_1}(\alpha) = -\frac{1}{N - k_0 + 1} \sum_{i \in \mathcal{NZ}_\alpha \setminus \{j\}} \widehat{\mu}_{n,k_1}(\alpha^{\oplus(j,i)}),$$

where $\alpha^{\oplus(j,i)} \in \widehat{\mathbb{F}}_2^{k_1 \times n}$ (for $i \neq j$) is defined as

$$(\alpha^{\oplus(j,i)})_\ell = \begin{cases} 0 & \text{if } \ell = j, \\ \alpha_i \oplus \alpha_j & \text{if } \ell = i, \\ \alpha_\ell & \text{if } \ell \notin \{i, j\}. \end{cases}$$

Note that $\#\alpha^{\oplus(j,i)} = k_0 - 1$ if $\alpha_i \oplus \alpha_j \neq 0$ and $\#\alpha^{\oplus(j,i)} = k_0 - 2$ if $\alpha_i \oplus \alpha_j = 0$ (i.e., $\alpha_i = \alpha_j$).

Proof. Denote $k_0 = k$. We assume that $\mathcal{NZ}_\alpha = [k_0] = [k]$, which is possible without loss of generality by Proposition 8.

We further assume that $k_1 = k$, as adding or removing zero elements from α does not change $\widehat{\mu}_{n,k}(\alpha)$. Finally, using Proposition 8 we assume without loss of

generality that $j = k$. By Proposition 2,

$$\begin{aligned}
\widehat{\mu}_{n,k}(\alpha) &= \mathbb{E}_{x \sim \mu_{n,k}} [\chi_\alpha(x)] = \mathbb{E}_{x \sim \mu_{n,k-1}} \left[\mathbb{E}_{x_k \sim \mathbb{F}_2^n \setminus \{x_1, \dots, x_{k-1}\}} [\chi_\alpha(x_{1..k-1}, x_k)] \right] \\
&= \frac{N}{N-k+1} \mathbb{E}_{x \sim \mu_{n,k-1}} \left[\mathbb{E}_{x_k \sim \mathbb{F}_2^n} [\chi_\alpha(x_{1..k-1}, x_k)] \right] \\
&\quad - \frac{k-1}{N-k+1} \mathbb{E}_{x \sim \mu_{n,k-1}} \left[\mathbb{E}_{x_k \sim \{x_1, \dots, x_{k-1}\}} [\chi_\alpha(x_{1..k-1}, x_k)] \right] \\
&= \frac{N}{N-k+1} \mathbb{E}_{x \sim \mu_{n,k-1}} [\chi_{\alpha_{1..k-1}}(x_{1..k-1})] \mathbb{E}_{x_k \sim \mathbb{F}_2^n} [\chi_{\alpha_k}(x_k)] \\
&\quad - \frac{1}{N-k+1} \sum_{i=1}^{k-1} \mathbb{E}_{x \sim \mu_{n,k-1}} [\chi_\alpha(x_{1..k-1}, x_i)] \\
&= 0 - \frac{1}{N-k+1} \sum_{i=1}^{k-1} \mathbb{E}_{x \sim \mu_{n,k-1}} [\chi_{(\alpha_{1..i-1}, \alpha_i \oplus \alpha_k, \alpha_{i+1..k-1})}(x_{1..k-1})] \\
&= -\frac{1}{N-k+1} \sum_{i=1}^{k-1} \mu_{n,k-1}(\alpha^{\oplus(k,i)}) = -\frac{1}{N-k+1} \sum_{i=1}^{k-1} \widehat{\mu}_{n,k}(\alpha^{\oplus(k,i)}),
\end{aligned}$$

where in the fifth equality we used $\mathbb{E}_{x_k \sim \mathbb{F}_2^n} [\chi_{\alpha_k}(x_k)] = \mathbb{E}[\chi_{\alpha_k} \chi_0] = 0$, which holds by orthogonality of characters since $\alpha_k \neq 0$. \blacksquare

Proposition 11 (Recursive bound for $\widehat{\mu}_{n,k_1}(\alpha)^2$). *For parameters $k_1 \geq k_0 \geq 2$, let $\alpha \in \widehat{\mathbb{F}}_2^{k_1 \times n}$ with $\#\alpha = k_0$, and define $\mathcal{NZ}_\alpha = \{i \in [k_1] \mid \alpha_i \neq 0\}$ ($|\mathcal{NZ}_\alpha| = k_0$). Then for any $j \in \mathcal{NZ}_\alpha$,*

$$\widehat{\mu}_{n,k_1}(\alpha)^2 \leq \frac{k_0 - 1}{(N - k_0 + 1)^2} \sum_{i \in \mathcal{NZ}_\alpha \setminus \{j\}} \widehat{\mu}_{n,k_1}(\alpha^{\oplus(j,i)})^2.$$

Proof. By Proposition 10 and the Cauchy–Schwarz inequality,

$$\begin{aligned}
\widehat{\mu}_{n,k_1}(\alpha)^2 &= \left(-\frac{1}{N-k_0+1} \sum_{i \in \mathcal{NZ}_\alpha \setminus \{j\}} \widehat{\mu}_{n,k_1}(\alpha^{\oplus(j,i)}) \right)^2 \\
&\leq \frac{k_0-1}{(N-k_0+1)^2} \sum_{i \in \mathcal{NZ}_\alpha \setminus \{j\}} \widehat{\mu}_{n,k_1}(\alpha^{\oplus(j,i)})^2.
\end{aligned}$$

\blacksquare

Lemma 1 ([13], Lemma 4.1 – Bound on magnitude of level- k Fourier coefficients). *Let $k_1 \geq k_0$ and $0 \leq k_0 \leq N/2$. Then, $M^{-k_0}[\mu_{n,k_1}] \leq \frac{1}{\sqrt{\binom{N}{k_0}}}$.*

A slightly stronger bound was also proved in Lemma 1 of [11], but we give the simpler proof of [13].

Proof. We may assume that $k_0 = k_1 = k$, as adding and removing 0 elements from α does not change $\widehat{\mu}_{n,k}(\alpha)$. The proof is by induction on k .

For $k = 0$, we have $M^{-0}[\mu_{n,k}] = 1 = \frac{1}{\sqrt{\binom{N}{0}}}$.

Next, let $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$ have $\#\alpha = k$. For $k = 1$, by Proposition 9, $|\widehat{\mu}_{n,k}(\alpha)| = 0 < \frac{1}{\sqrt{\binom{N}{1}}}$. For $k \geq 2$, by Proposition 10 and the triangle inequality,

$$|\widehat{\mu}_{n,k}(\alpha)| = \left| -\frac{1}{N-k+1} \sum_{i=1}^{k-1} \widehat{\mu}_{n,k}(\alpha^{\oplus(k,i)}) \right| \leq \frac{1}{N-k+1} \sum_{i=1}^{k-1} |\widehat{\mu}_{n,k}(\alpha^{\oplus(k,i)})|.$$

We have $\#\alpha^{\oplus(k,i)} \in \{k-1, k-2\}$. Assume that for m values of $i \in [k-1]$, $\#\alpha^{\oplus(k,i)} = k-2$ holds. Then, by the induction hypothesis (assuming $k \leq N/2$),

$$\begin{aligned} |\widehat{\mu}_{n,k}(\alpha)| &\leq \frac{m}{N-k+1} M^{=k-2}[\mu_{n,k}] + \frac{k-1-m}{N-k+1} M^{=k-1}[\mu_{n,k}] \\ &\leq \frac{m}{N-k+1} \frac{1}{\sqrt{\binom{N}{k-2}}} + \frac{k-1-m}{N-k+1} \frac{1}{\sqrt{\binom{N}{k-1}}} \leq \frac{k-1}{N-k+1} \frac{1}{\sqrt{\binom{N}{k-2}}} \\ &= \frac{k-1}{N-k+1} \sqrt{\frac{k-2}{N} \frac{k-3}{N-1} \cdots \frac{1}{N-(k-3)}} \\ &\leq \sqrt{\frac{k}{N-k+2} \frac{k-1}{N-k+1} \sqrt{\frac{k-2}{N} \frac{k-3}{N-1} \cdots \frac{1}{N-(k-3)}}} = \frac{1}{\sqrt{\binom{N}{k}}}. \end{aligned}$$

■

Lemma 2 ([11], Lemma 2 – Bound on level- k Fourier weight).

$$\text{For } 1 \leq k \leq \frac{N}{2}, W^{=k}[\mu_{n,k}] \leq \left(\frac{k}{N-k} \right)^{k/2}.$$

Proposition 12. Let $k_1 \geq k_0$ and $2 \leq k_0 < N/2$ for k even. Let $\alpha \in \widehat{\mathbb{F}}_2^{k_1 \times n}$ have $\#\alpha = k_0$. Assume that $\alpha_i = \alpha_j$ for all $i, j \in [k_1]$ such that $\alpha_i, \alpha_j \neq 0$ (i.e., $i, j \in \mathcal{NZ}_\alpha$). Then,

$$\widehat{\mu}_{n,k_1}(\alpha) = (-1)^{k_0/2} \frac{k_0-1}{N-1} \frac{k_0-3}{N-3} \cdots \frac{1}{N-(k_0-1)}.$$

Moreover,

$$\frac{1}{\sqrt{k \binom{N}{k_0}}} \leq |\widehat{\mu}_{n,k_1}(\alpha)| \leq \frac{1}{\sqrt{\binom{N}{k_0}}}.$$

Proof. We assume without loss of generality that $k_0 = k_1 = k$. The proof is by induction on k . By Proposition 10,

$$\widehat{\mu}_{n,k}(\alpha) = -\frac{1}{N-k+1} \sum_{i=1}^{k-1} \widehat{\mu}_{n,k}(\alpha^{\oplus(k,i)}).$$

For $k = 2$, this gives $-\frac{1}{N-1} \widehat{\mu}_{n,2}(\alpha^{\oplus(2,1)}) = -\frac{1}{N-1}$, as $\#\alpha^{\oplus(2,1)} = 0$ and hence $\widehat{\mu}_{n,2}(\alpha^{\oplus(2,1)}) = 1$.

For $k > 2$, for all $i \in [k-1]$, $\alpha^{\oplus(k,i)}$ is equal to $\alpha^{\oplus(k,1)}$ (up to a permutation of the elements). Therefore, $\widehat{\mu}_{n,k}(\alpha) = -\frac{k-1}{N-k+1} \widehat{\mu}_{n,k}(\alpha^{\oplus(k,1)})$. Since $\#\alpha^{\oplus(k,1)} =$

$k-2$, and $\alpha^{\oplus(k,1)}$ has all non-zero elements equal (as α), we apply the induction hypothesis to $\alpha^{\oplus(k,1)}$ and deduce

$$\widehat{\mu}_{n,k}(\alpha) = -\frac{k-1}{N-k+1}(-1)^{k/2-1}\frac{k-3}{N-1}\cdots\frac{1}{N-(k-3)} = (-1)^{k/2}\frac{k-1}{N-1}\frac{k-3}{N-3}\cdots\frac{1}{N-(k-1)}.$$

Next, note that $|\widehat{\mu}_{n,k}(\alpha)| \leq \frac{1}{\sqrt{\binom{N}{k}}}$ holds by Lemma 1. It remains to prove that $\frac{1}{\sqrt{k\binom{N}{k}}} \leq |\widehat{\mu}_{n,k}(\alpha)|$. Indeed,

$$\begin{aligned} \sqrt{k}|\widehat{\mu}_{n,k}(\alpha)| &= \sqrt{k}\frac{k-1}{N-1}\frac{k-3}{N-3}\cdots\frac{1}{N-(k-1)} \geq \sqrt{\frac{k}{N-(k-1)}}\sqrt{\frac{k-1}{N-1}\frac{k-2}{N-2}\cdots\frac{1}{N-(k-1)}} \\ &\geq \sqrt{\frac{k}{N}}\sqrt{\frac{k-1}{N-1}\frac{k-2}{N-2}\cdots\frac{1}{N-(k-1)}} = \frac{1}{\sqrt{\binom{N}{k}}}. \end{aligned}$$

■

3 Framework for Bounding the Weight of $\widehat{\mu}_{n,k}$ on Structured Subsets

We begin with a motivating example for our framework.

Let $r \geq 2$ and $k \geq 1$ be parameters such that k is not too large compared to N . Suppose we want to upper bound the expression

$$\sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{k \times n} \\ \#\alpha = k}} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 = \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2,$$

where $\alpha^{\odot r} = (\alpha_1^{\odot r}, \dots, \alpha_k^{\odot r})$. Since $\#\alpha^{\odot r} = r\#\alpha = rk$, we apply Lemma 1 and obtain

$$\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 \leq N^k M^{-(rk)}[\mu_{n,rk}] \leq N^k \frac{1}{\binom{N}{rk}}. \quad (3)$$

Another option to bound the expression is to use Lemma 2 and deduce

$$\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 \leq \sum_{\beta \in \mathcal{M}_{=rk,rk}^n} \widehat{\mu}_{n,rk}(\beta)^2 = W^{(rk)}[\mu_{n,rk}] \leq \left(\frac{rk}{N-rk}\right)^{rk/2}. \quad (4)$$

The bounds obtained above are far from tight in general, as they make little use of the structure of the subset of masks we sum over. In order to improve the bound for $r \geq 3$, for every $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$ with $\#\alpha = k$ apply Proposition 11 to $\alpha^{\odot r}$ (with $k_1 = k_0 = rk$ and $j = rk$), obtaining

$$\begin{aligned} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 &\leq \frac{rk-1}{(N-rk+1)^2} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \sum_{i=1}^{rk-1} \widehat{\mu}_{n,rk}((\alpha^{\odot r})^{\oplus(rk,i)})^2 \\ &= \frac{rk-1}{(N-rk+1)^2} \sum_{i=1}^{rk-1} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}((\alpha^{\odot r})^{\oplus(rk,i)})^2. \end{aligned} \quad (5)$$

Fix $i \in [rk - 1]$. We now analyze the term $\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}((\alpha^{\odot r})^{\oplus(rk,i)})^2$. A crucial observation in the analysis is that since $r \geq 3$ and $(\alpha^{\odot r})^{\oplus(rk,i)}$ changes only indices rk and i of $\alpha^{\odot r}$, then for any $i \in [rk - 1]$, $(\alpha^{\odot r})^{\oplus(rk,i)}$ fully determines α (and $\alpha^{\odot r}$). Indeed, since $r \geq 3$, for every $\ell \in [k]$, α_ℓ still appears in at least one entry of $(\alpha^{\odot r})^{\oplus(rk,i)}$ (for every $i \in [rk - 1]$). This does not hold for $r = 2$ and $i = 2k - 1$ (as $(\alpha^{\odot 2})^{\oplus(2k,2k-1)}$ is independent of α_k).

In other words, given $i \in [rk - 1]$ the i 'th operation in Proposition 11 applied to $\alpha^{\odot r}$ (whose outcome is $(\alpha^{\odot r})^{\oplus(rk,i)}$) is invertible for $r \geq 3$. Since $\#(\alpha^{\odot r})^{\oplus(rk,i)} \in \{rk - 1, rk - 2\}$,

$$\begin{aligned} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}((\alpha^{\odot r})^{\oplus(rk,i)})^2 &\leq \sum_{\beta \in \mathcal{M}_{=rk-1,rk-1}^n} \widehat{\mu}_{n,rk-1}(\beta)^2 + \sum_{\beta \in \mathcal{M}_{=rk-2,rk-2}^n} \widehat{\mu}_{n,rk-2}(\beta)^2 \\ &= W^{=rk-1}[\mu_{n,rk-1}] + W^{=rk-2}[\mu_{n,rk-2}], \end{aligned}$$

where the inequality crucially uses the fact that for every $i \in [rk - 1]$, in the (multi) set $\{(\alpha^{\odot r})^{\oplus(rk,i)} \mid \alpha \in \widehat{\mathbb{F}}_2^{k \times n} \wedge \#\alpha = k\}$ each mask appears only once. This holds due to the invertibility of $(\alpha^{\odot r})^{\oplus(rk,i)}$. Combining with (5),

$$\begin{aligned} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 &\leq \frac{rk-1}{(N-rk+1)^2} \sum_{i=1}^{rk-1} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}((\alpha^{\odot r})^{\oplus(rk,i)})^2 \\ &\leq \frac{rk-1}{(N-rk+1)^2} \sum_{i=1}^{rk-1} (W^{=rk-1}[\mu_{n,rk-1}] + W^{=rk-2}[\mu_{n,rk-2}]) \\ &= \left(\frac{rk-1}{N-rk+1}\right)^2 (W^{=rk-1}[\mu_{n,rk-1}] + W^{=rk-2}[\mu_{n,rk-2}]). \end{aligned}$$

We can now use Lemma 2 to bound $W^{=rk-1}[\mu_{n,rk-1}] + W^{=rk-2}[\mu_{n,rk-2}]$, and obtain a significant improvement over (4) due to the multiplication by $\left(\frac{rk-1}{N-rk+1}\right)^2$.

In general, there is no reason to stop after one application Proposition 11. We can obtain improved bounds by applying Proposition 11 recursively to each mask in each set $\{(\alpha^{\odot r})^{\oplus(rk,i)} \mid \alpha \in \widehat{\mathbb{F}}_2^{k \times n} \mid \#\alpha = k\}$ for $i \in [rk - 1]$. The outcome is a recursion tree and we apply Lemma 2 only at the leaves.

We now describe our framework which allows to obtain bounds on the Fourier weight of $\mu_{n,k}$ on structured subsets of masks, generalizing the above analysis.

3.1 General Framework

We consider the following initial setting.

Setting 1 Let $k' > 0$ be an integer parameter. Let \mathcal{S} be a set of strings. Let $T : \mathcal{S} \rightarrow \widehat{\mathbb{F}}_2^{k' \times n}$ be a mapping such that the following two restrictions hold:

- (a1) T is injective on the elements of \mathcal{S} , and
- (a2) there is a non-zero index subset $\mathcal{NZ} \subseteq [k']$ such that for every $\alpha \in \mathcal{S}$ and every $\ell \in [k']$, $T(\alpha)_\ell \neq 0$ if and only if $\ell \in \mathcal{NZ}$.

The second restriction implies that all masks in $\{T(\alpha) \mid \alpha \in \mathcal{S}\}$ have level $\#T(\alpha) = |\mathcal{NZ}|$.

Assume that our goal is to bound $\sum_{\alpha \in \mathcal{S}} \widehat{\mu}_{n,k'}(T(\alpha))^2$. Note that $\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2$ is a special case with $k' = rk$, $\mathcal{S} = \mathcal{M}_{=k,k}^n$ and $T(\alpha) = T_{r,k}(\alpha) = \alpha^{\odot r}$, where α is duplicated r times (here $\mathcal{NZ} = [rk]$).

We start from the initial set of masks $\{T(\alpha) \mid \alpha \in \mathcal{S}\}$, and invoke recursive calls of Proposition 11, where Lemma 2 is applied only at the leaves of the recursion tree.

Let $\beta \in \widehat{\mathbb{F}}_2^{k' \times n}$ be a mask. To formally define the recursion, consider the operation $\beta^{\oplus(j,i)}$ for $j \in \mathcal{NZ} = \mathcal{NZ}_\beta$ and $i \in \mathcal{NZ} \setminus \{j\}$. The formula of Proposition 11 applied to β includes $|\mathcal{NZ}| - 1$ such operations, where j is fixed and i ranges over all $\mathcal{NZ} \setminus \{j\}$. Thus, we call index j the *primary index*, while we call each $i \in \mathcal{NZ} \setminus \{j\}$ a *secondary index*.

Each recursive node v at depth $d \geq 0$ is labeled by a recursion stack, which consists of the sequence of d secondary indices $i_1, \dots, i_d \in [k']$ for the recursive calls up to this node, and a sequence of bits $b_1, \dots, b_d \in \{0, 1\}$. For $d' \in [d]$, bit $b_{d'}$ specifies whether the outcome of the XOR operation at index $i_{d'}$ was zero or not. The purpose of these bits is to keep track of the set \mathcal{NZ} that evolves during the recursion.

We will assume that there is a *primary index selector*, or PIS, which is an application-dependent procedure that selects the next primary index (denoted j_{d+1}) for the invocation of Proposition 11. The input to the PIS includes the recursion stack $v = (i_1, \dots, i_d, b_1, \dots, b_d)$. Initially, the recursion stack is empty, and thus the first primary index should be fixed. For example, for $T(\alpha) = \alpha^{\odot r}$ and we initially simply set $j_1 = rk$. We remark that the PIS also depends on the initial parameters of Setting 1, (\mathcal{S}, T) . However, (\mathcal{S}, T) are assumed to be fixed and hardcoded inside the PIS.

Fixing a PIS implementation pis , we define a recursive procedure up to depth d_{\max} (called $\text{calcW}_{pis, d_{\max}}$) for upper bounding the weight $\sum_{\alpha \in \mathcal{S}} \widehat{\mu}_{n,k'}(T(\alpha))^2$.

Definition of calcW. The procedure $\text{calcW}_{pis, d_{\max}}$ obtains 5 parameters:

- (1) (current) recursion depth d ,
- (2) stack trace $v = (i_1, \dots, i_d, b_1, \dots, b_d)$,
- (3) set \mathcal{S}_v ,
- (4) mapping $T_v : \mathcal{S}_v \rightarrow \widehat{\mathbb{F}}_2^{k' \times n}$, and
- (5) set $\mathcal{NZ}_v \subset [k']$ such that for all $\alpha \in \mathcal{S}_v$, $T_v(\alpha)_i \neq 0$ if and only if $i \in \mathcal{NZ}_v$.

Initially, \mathcal{S}, T are defined by Setting 1, and thus $d = 0$, $v = \text{NULL}$, $\mathcal{S}_v = \mathcal{S}$, $T_v = T$ and $\mathcal{NZ}_v = \mathcal{NZ}$. In most (but not all) of our applications, $\mathcal{NZ}_v = [k']$, as the level of all masks $T(\alpha)$ for $\alpha \in \mathcal{S}$ will be k' .

$\text{calcW}_{pis, d_{\max}}(d, v = (i_1, \dots, i_d, b_1, \dots, b_d), \mathcal{S}_v, T_v, \mathcal{NZ}_v)$

1. $k'_v \leftarrow |\mathcal{NZ}_v|$.
2. If $d = d_{\max}$, return $(\frac{k'_v}{N - k'_v})^{k'_v/2}$.

3. $j \leftarrow \text{pis}(v)$.
4. $W \leftarrow 0$.
5. For all $i \in \mathcal{NZ}_v \setminus \{j\}$:
 - (a) $v_{i,0} \leftarrow (i_1, \dots, i_d, i, b_1, \dots, b_d, 0)$, $v_{i,1} \leftarrow (i_1, \dots, i_d, i, b_1, \dots, b_d, 1)$.
 - (b) Define $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ for every $\alpha \in \mathcal{S}_v$.
 - (c) $T_{v_{i,0}} \leftarrow T_{v_i}$, $T_{v_{i,1}} \leftarrow T_{v_i}$.
 - (d) $\mathcal{S}_{v_{i,0}} \leftarrow \{\alpha \in \mathcal{S}_v \mid T_{v_i}(\alpha)_i = 0\}$, $\mathcal{S}_{v_{i,1}} \leftarrow \{\alpha \in \mathcal{S}_v \mid T_{v_i}(\alpha)_i \neq 0\}$.
 - (e) $\mathcal{NZ}_{v_{i,0}} \leftarrow \mathcal{NZ}_v \setminus \{i, j\}$, $\mathcal{NZ}_{v_{i,1}} \leftarrow \mathcal{NZ}_v \setminus \{j\}$.
 - (f) $W \leftarrow W + \text{calcW}_{\text{pis}, d_{\max}}(d+1, v_{i,0}, \mathcal{S}_{v_{i,0}}, T_{v_{i,0}}, \mathcal{NZ}_{v_{i,0}})$,
 $W \leftarrow W + \text{calcW}_{\text{pis}, d_{\max}}(d+1, v_{i,1}, \mathcal{S}_{v_{i,1}}, T_{v_{i,1}}, \mathcal{NZ}_{v_{i,1}})$.
6. Return $\frac{k'_v - 1}{(N - k'_v + 1)^2} W$.

Remark 2. Assume that \mathcal{S}, T and pis are fixed. Since the output of pis only depends on the recursion stack (but not on specific masks), the primary indices are uniquely defined by the recursion stack v , even though v does not include them explicitly. More generally, the 4 parameters $d, \mathcal{S}_v, T_v, \mathcal{NZ}_v$ of calcW are uniquely determined by v . The only reason we explicitly include them as parameters of calcW is to simplify its description.

Applicability of calcW . The correctness of calcW will rely on the assumption that the two restrictions of Setting 1 hold at all nodes, as they will be crucial for applying Lemma 2 at the leaves. Since the variable \mathcal{NZ}_v is updated correctly, restriction (a2) indeed holds at all nodes. However, restriction (a1) may not hold recursively, and it requires special treatment depending on the specific application. We formalize the corresponding conditions in the following definition.

Definition 6 (Applicability of calcW). *We say that calcW is applicable up to depth d_{\max} with parameters (\mathcal{S}, T) and a PIS pis , if the following conditions hold:*

- (b1) *the pair (\mathcal{S}, T) satisfies the restrictions of Setting 1, and*
- (b2) *considering the recursion tree with root $\text{calcW}_{\text{pis}, d_{\max}}(0, (NULL), \mathcal{S}, T, \mathcal{NZ})$ (\mathcal{NZ} is defined in Setting 1): for every node v at depth at most d_{\max} such that $j = \text{pis}(v)$, for all $i \in \mathcal{NZ}_v \setminus \{j\}$ and $\alpha \in \mathcal{S}_v$, $T_v(\alpha)$ can be (uniquely) recovered from $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$.*

Before formally analyzing calcW , we simplify the second condition of Definition 6. This simplification will be useful in applications.

Proposition 13 (Sufficient condition for applicability of calcW). *Given (\mathcal{S}, T) and a PIS pis , assume that*

- (c1) *the pair (\mathcal{S}, T) satisfies the restrictions of Setting 1, and*
- (c2) *considering the recursion tree with root $\text{calcW}_{\text{pis}, d_{\max}}(0, (NULL), \mathcal{S}, T, \mathcal{NZ})$: for every node v at depth at most d_{\max} such that $j = \text{pis}(v)$, for all $i \in \mathcal{NZ}_v \setminus \{j\}$ and $\alpha \in \mathcal{S}_v$, $T_v(\alpha)_j$ can be recovered from $T_{v_i}(\alpha)$.*

Then, $\text{calcW}_{d_{\max}}$ is applicable up to depth d_{\max} with \mathcal{S}, T and pis .

Proof. Condition **(b1)** of Definition 6 holds by assumption. We prove condition **(b2)**. Fix a node v of depth at most d_{\max} and let $\alpha \in \mathcal{S}_v$. According to Definition 6, we need to prove that $T_v(\alpha)$ can be recovered from $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ (for all $i \in \mathcal{NZ}_v \setminus \{j\}$). Since only indices j and i are modified in $T_v(\alpha)$ by the mapping T_{v_i} , it is sufficient to prove that both $T_v(\alpha)_j$ and $T_v(\alpha)_i$ can be computed from $T_{v_i}(\alpha)$. By assumption, $T_v(\alpha)_j$ can be recovered from $T_{v_i}(\alpha)$. Moreover, since $T_{v_i}(\alpha)_i = (T_v(\alpha)^{\oplus(j,i)})_i = T_v(\alpha)_i \oplus T_v(\alpha)_j$, then $T_v(\alpha)_i = T_{v_i}(\alpha)_i \oplus T_v(\alpha)_j$ can also be recovered from $T_{v_i}(\alpha)$. Hence both conditions of Definition 6 hold. ■

The following definition will be useful in applications.

Definition 7 (Unaltered index). An index $\ell' \in [k']$ is called *unaltered* at a node $v = (i_1, \dots, i_d, b_1, \dots, b_d)$ if ℓ' has not been selected as primary or secondary index. Namely, $\ell' \neq j_{d'}$ and $\ell' \neq i_{d'}$ for all $d' \in [d]$.

The definition is motivated by the simple property that if ℓ' is unaltered at node v , then for any $\alpha \in \mathcal{S}_v$, $T_v(\alpha)_{\ell'} = T(\alpha)_{\ell'}$ (where T is the initial mapping at the root). This property holds since the mappings T_{v_i} at any node v only modify the entries of the primary index j and secondary index i .

Denote by \mathcal{U}_v the set of all unaltered indices at node v . At the root node v , $\mathcal{U}_v = [k']$. Since every child of any node v has one primary and one secondary index, a node at depth d has $|\mathcal{U}_v| \geq k' - 2d$.

Analysis of calcW. We now analyze calcW, assuming it is applicable up to a certain depth according to Definition 6.

Proposition 14 (Recursive validity of Setting 1). Assume that calcW is applicable up to depth d_{\max} with parameters \mathcal{S} , T and a PIS, *pis*. Then, for each node v at depth at most d_{\max} , **(d1)** T_v is injective on the elements of \mathcal{S}_v , **(d2)** for every $\alpha \in \mathcal{S}_v$ and every $\ell \in [k']$, $T_v(\alpha)_\ell \neq 0$ if and only if $\ell \in \mathcal{NZ}_v$.

Proof. The proof is by induction on the depth $d \leq d_{\max}$ of v . The two restrictions hold at the root ($d = 0$) by assumption. Assume correctness up to depth d and let v be a node of depth d . Consider a child node $v_{i,b}$ for $i \in \mathcal{NZ}_v \setminus \{j\}$ and $b \in \{0, 1\}$. Recall that $T_{v_{i,b}}(\alpha) = T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ only changes entries i, j of $T_v(\alpha)$.

We prove **(d1)**. Consider $\alpha, \beta \in \mathcal{S}_{v_{i,b}}$ such that $T_{v_{i,b}}(\alpha) = T_{v_{i,b}}(\beta)$. We show that $\alpha = \beta$. Since calcW is applicable up to depth d_{\max} , condition **(b2)** of Definition 6 implies that $T_v(\alpha) = T_v(\beta)$. Indeed, if $T_v(\alpha) \neq T_v(\beta)$ but $T_{v_{i,b}}(\alpha) = T_{v_{i,b}}(\beta)$ then $T_v(\alpha)$ cannot be uniquely recovered from $T_{v_{i,b}}(\alpha) = T_{v_i}(\alpha)$.

Since $T_v(\alpha) = T_v(\beta)$, the induction hypothesis implies that $\alpha = \beta$ (as $\mathcal{S}_{v_{i,b}} \subseteq \mathcal{S}_v$ and T_v is injective of \mathcal{S}_v). This proves **(d1)**.

We prove **(d2)**. Consider $\alpha \in \mathcal{S}_{v_{i,b}}$ and let $\ell \in [k']$. If $\ell = j$, then $T_{v_{i,b}}(\alpha)_\ell = 0$ and $\ell \notin \mathcal{NZ}_{v_{i,b}}$ by definition of calcW.

Next, consider $\ell = i$. Then $T_v(\alpha)_\ell \neq 0$ and $\ell \in \mathcal{NZ}_v$ by the hypothesis. Therefore, if $T_{v_{i,b}}(\alpha)_\ell \neq 0$, then $b = 1$ and also $i \in \mathcal{NZ}_{v_{i,1}}$, while if $T_{v_{i,b}}(\alpha)_\ell = 0$, then $b = 0$ and also $i \notin \mathcal{NZ}_{v_{i,0}}$ (by definition of calcW).

Otherwise $\ell \notin \{i, j\}$. Then $T_{v_{i,b}}(\alpha)_\ell = T_v(\alpha)_\ell$, so $T_{v_{i,b}}(\alpha)_\ell \neq 0$ if and only if $T_v(\alpha)_\ell \neq 0$. By the induction hypothesis, this holds if and only if $\ell \in \mathcal{NZ}_v$, which holds if and only if $\ell \in \mathcal{NZ}_{v_{i,b}}$ (by definition of calcW). This completes the proof. \blacksquare

Proposition 15 (Correctness of calcW). *Assume that calcW is applicable up to depth $d_{\max} \geq 0$ with parameters (\mathcal{S}, T) and a PIS, pis . Then, for every node v with depth $d \leq d_{\max}$ such that $|\mathcal{NZ}_v| \leq \frac{N}{2}$,*

$$\sum_{\alpha \in \mathcal{S}_v} \widehat{\mu}_{n,k'}(T_v(\alpha))^2 \leq \text{calcW}_{\text{pis}, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v).$$

Proof. We prove the result by induction on $d \leq d_{\max}$ (starting with $d = d_{\max}$, down to $d = 0$). Let $k'_v = |\mathcal{NZ}_v|$. For $d = d_{\max}$,

$$\begin{aligned} \sum_{\alpha \in \mathcal{S}_v} \widehat{\mu}_{n,k'}(T_v(\alpha))^2 &\leq \sum_{\beta \in \widehat{\mathbb{F}}_2^{k'_v}} \widehat{\mu}_{n,k'}(\beta)^2 = W^{=k'_v}[\mu_{n,k'_v}] \leq \left(\frac{k'_v}{N-k'_v}\right)^{k'_v/2} \\ &= \text{calcW}_{\text{pis}, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v), \end{aligned}$$

where the first inequality relies on **(d2)** in Proposition 14, as we delete the $k' - |\mathcal{NZ}_v|$ zero entries that are common to all $T_v(\alpha)$ for $\alpha \in \mathcal{S}_v$. It further relies on **(d1)** in Proposition 14, as each $\alpha \in \mathcal{S}_v$ is mapped to a single $\beta \in \widehat{\mathbb{F}}_2^{k'_v}$ after removing the common zero entries. The second inequality is by Lemma 2.

For $d < d_{\max}$, by reordering elements, we assume without loss of generality that $\mathcal{NZ}_v = [k'_v]$ and $\text{pis}(v) = k'_v$. Then, by Proposition 11,

$$\begin{aligned} \sum_{\alpha \in \mathcal{S}_v} \widehat{\mu}_{n,k'}(T_v(\alpha))^2 &\leq \frac{k'_v-1}{(N-k'_v+1)^2} \sum_{\alpha \in \mathcal{S}_v} \sum_{i=1}^{k'_v-1} \widehat{\mu}_{n,k'}(T_v(\alpha)^{\oplus(k'_v, i)})^2 \\ &= \frac{k'_v-1}{(N-k'_v+1)^2} \sum_{i=1}^{k'_v-1} \left(\sum_{\alpha \in \mathcal{S}_{v_{i,0}}} \widehat{\mu}_{n,k'}(T_v(\alpha)^{\oplus(k'_v, i)})^2 + \sum_{\alpha \in \mathcal{S}_{v_{i,1}}} \widehat{\mu}_{n,k'}(T_v(\alpha)^{\oplus(k'_v, i)})^2 \right), \end{aligned} \tag{6}$$

where we use the fact that $\mathcal{S}_{v_{i,0}} \cup \mathcal{S}_{v_{i,1}} = \mathcal{S}_v$ for every $i \in [k'_v - 1]$. We have

$$\begin{aligned} \sum_{\alpha \in \mathcal{S}_{v_{i,0}}} \widehat{\mu}_{n,k'}(T_v(\alpha)^{\oplus(k'_v, i)})^2 &= \sum_{\alpha \in \mathcal{S}_{v_{i,0}}} \widehat{\mu}_{n,k'}(T_{v_{i,0}}(\alpha))^2 \\ &\leq \text{calcW}_{\text{pis}, d_{\max}}(d+1, v_{i,0}, \mathcal{S}_{v_{i,0}}, T_{v_{i,0}}, \mathcal{NZ}_{v_{i,0}}), \end{aligned}$$

where the inequality is by the induction hypothesis (relying on applicability up to depth d_{\max}).

Moreover, a similar inequality holds for the sum over $\mathcal{S}_{v_{i,1}}$. Plugging these inequalities into (6), and comparing with the return value of calcW , we deduce $\sum_{\alpha \in \mathcal{S}_v} \widehat{\mu}_{n,k'}(T_v(\alpha))^2 \leq \text{calcW}_{\text{pis}, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v)$, concluding the proof. \blacksquare

Proposition 16. *Let v be a node of depth d such that $d \leq d_{\max}$ and $k'_v = |\mathcal{NZ}_v| \leq N/8$. Denote $d' = d_{\max} - d$. Then,*

$$\text{calcW}_{pis, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v) \leq 2^{d'} \frac{(k'_v)^{2d'} (k'_v - 2d')^{k'_v/2 - d'}}{(N - k'_v)^{k'_v/2 + d'}}.$$

Proof. To simplify notation, denote $k = k'_v$. The recursion tree starting from v is of depth $d' = d_{\max} - d$. Each leaf u contributes to the output at most

$$\left(\frac{k}{N-k}\right)^{d'} \left(\frac{k'_u}{N-k'_u}\right)^{k'_u/2}, \quad (7)$$

where we used the fact that for each internal node w , $k'_w \leq k$ and thus $\frac{k'_w - 1}{(N - k'_w + 1)^2} \leq \frac{k-1}{(N-k+1)^2} \leq \frac{k}{(N-k)^2}$.

Initially, $|\mathcal{NZ}_v| = k'_v = k$. For each internal node w , for each $i \in \mathcal{NZ}_w \setminus \{j\}$, $|\mathcal{NZ}_{w_{i,0}}| = |\mathcal{NZ}_w| - 2$ (there are $k'_w < k$ such children $w_{i,0}$) and $|\mathcal{NZ}_{w_{i,1}}| = |\mathcal{NZ}_w| - 1$ (there are $k'_w < k$ such children $w_{i,1}$).

Therefore, for every leaf u , $k'_u \in [k - d', k - 2d']$. More specifically for $c \in \{0, 1, \dots, d'\}$, the number of leaf nodes u with $k'_u = k - 2d' + c$ is at most $k^{d'} \binom{d'}{c}$. Hence, using (7), we bound

$$\begin{aligned} \text{calcW}_{pis, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v) &\leq \left(\frac{k}{N-k}\right)^{d'} \sum_{u \text{ leaf}} \left(\frac{k'_u}{N-k'_u}\right)^{k'_u/2} \\ &\leq \left(\frac{k}{N-k}\right)^{d'} k^{d'} \sum_{c=0}^{d'} \binom{d'}{c} \left(\frac{k-2d'+c}{N-k+2d'-c}\right)^{(k-2d'+c)/2} \\ &\leq \left(\frac{k}{N-k}\right)^{2d'} 2^{d'} \max_{c \in \{0, \dots, d'\}} \left\{ \left(\frac{k-2d'+c}{N-k+2d'-c}\right)^{(k-2d'+c)/2} \right\}. \end{aligned} \quad (8)$$

Denote $B(c) = \left(\frac{k-2d'+c}{N-k+2d'-c}\right)^{(k-2d'+c)/2}$. For $c+1 \leq d'$, by Proposition 1,

$$\begin{aligned} \frac{B(c+1)}{B(c)} &\leq e^{\frac{(k-2d'+c)/2}{k-2d'+c} + \frac{(k-2d'+c)/2}{N-k+2d'-c-1}} \left(\frac{k-2d'+c+1}{N-k+2d'-c-1}\right)^{1/2} \\ &\leq e^{\frac{1}{2} + \frac{k/2}{N-k}} \left(\frac{k}{N-k}\right)^{1/2} \leq e^{4/7} \left(\frac{1}{7}\right)^{1/2} \leq 1, \end{aligned}$$

where we have used the assumption that $k \leq N/8$. Thus,

$$\begin{aligned} \max_{c \in \{0, \dots, d'\}} \left\{ \left(\frac{k-2d'+c}{N-k+2d'-c}\right)^{(k-2d'+c)/2} \right\} &= \max_{c \in \{0, \dots, d'\}} \{B(c)\} = B(0) \\ &= \left(\frac{k-2d'}{N-k+2d'}\right)^{(k-2d')/2} \leq \left(\frac{k-2d'}{N-k}\right)^{(k-2d')/2}. \end{aligned}$$

Finally, plugging this back into (8) we deduce

$$\begin{aligned} \text{calcW}_{pis, d_{\max}}(d, v, \mathcal{S}_v, T_v, \mathcal{NZ}_v) &\leq \left(\frac{k}{N-k}\right)^{2d'} 2^{d'} \max_{c \in \{0, \dots, d'\}} \{B(c)\} \\ &\leq 2^{d'} \left(\frac{k}{N-k}\right)^{2d'} \left(\frac{k-2d'}{N-k}\right)^{(k-2d')/2} = 2^{d'} \frac{k^{2d'} (k-2d')^{(k-2d')/2}}{(N-k)^{k/2 + d'}}. \end{aligned}$$

■

Lemma 3. Assume that calcW is applicable up to depth $d = d_{\max} \geq 0$ with parameters (\mathcal{S}, T) and a PIS, pis . Assume further that initially $\#T(\alpha) = k_0 \leq N/8$ for all $\alpha \in \mathcal{S}$. Then,

$$\sum_{\alpha \in \mathcal{S}} \widehat{\mu}_{n, k_0}(T(\alpha))^2 \leq 2^d \frac{(k_0)^{2d} (k_0 - 2d)^{k_0/2-d}}{(N - k_0)^{k_0/2+d}} \leq 2^d \left(\frac{k_0}{N - k_0} \right)^{k_0/2+d}.$$

Ignoring the (relatively minor) term 2^d , the improvement over the naive application of Lemma 2 is by a factor of $(\frac{k_0}{N - k_0})^d$. This emphasizes the importance of defining a PIS that allows applying calcW up to a large depth d .

Proof. Let \mathcal{NZ} be that set defined in Setting 1. By Proposition 15 and Proposition 16,

$$\sum_{\alpha \in \mathcal{S}} \widehat{\mu}_{n, k'}(T(\alpha))^2 \leq \text{calcW}_{\text{pis}, d}(0, (\text{NULL}), \mathcal{S}, T, \mathcal{NZ}) \leq 2^d \frac{(k')^{2d} (k' - 2d)^{k'/2-d}}{(N - k')^{k'/2+d}}.$$

■

Remark 3. There are many possible variants of calcW that may give better bounds in different settings, but are not used in this paper. We summarize a few below.

1. Instead of fixing the maximal depth d_{\max} in advance, we can continue recursive calls from a node v as long as condition **(c2)** of Proposition 13 holds.
2. The purpose of condition **(b2)** of Definition 6 (or condition **(c2)** of Proposition 13) is to assure that T_v remains injective on the elements of \mathcal{S}_v at all nodes v . This can be assured without this condition if we partition \mathcal{S}_v into more subsets that result in more recursive calls (with additional information about the masks added to the recursion stack v to assure injectivity).
3. Instead of using the bound derived from Lemma 2, $(\frac{k'_v}{N - k'_v})^{k'_v/2}$, at the leaves with $d = d_{\max}$, we can use a bound derived from Lemma 1 (or a minimum of these bounds).

4 Indistinguishability Upper and Lower Bounds for SXoP $[r, n]$

In this section, we analyze the SXoP $[r, n]$ construction, proving the main theorem below.

Theorem 1. Assume that $rq \leq N/8$ and $N \geq 2^{13}r$. The following bounds hold depending on r .

Odd $r \geq 3$.

$$\begin{aligned} \text{Opt}_{\text{SXoP}[r, n]}^{\text{prf}}(q) &\leq 2^{r-1} r^r \frac{q}{N^{r-0.5}} \leq O_r \left(\frac{q}{N^{r-0.5}} \right), \text{ and} \\ \text{Opt}_{\text{SXoP}[r, n], u}^{\text{mu-prf}}(q_{\max}) &\leq 2^{r-0.5} r^r \frac{\sqrt{u} q_{\max}}{N^{r-0.5}} \leq O_r \left(\frac{\sqrt{u} q_{\max}}{N^{r-0.5}} \right), \end{aligned} \tag{9}$$

where the second inequality also requires $2^{r-0.5} r^r \frac{\sqrt{u} q_{\max}}{N^{r-0.5}} \leq \frac{1}{2}$.

$r = 2$.

$$\text{Opt}_{\text{SXoP}[2,n]}^{\text{prf}}(q) \leq \frac{5q}{N} \leq O\left(\frac{q}{N}\right). \quad (10)$$

Even $r \geq 4$.

$$\begin{aligned} \text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) &\leq 2r^{r/2} \frac{q}{N^{r/2}} \leq O_r\left(\frac{q}{N^{r/2}}\right), \text{ and} \\ \text{Opt}_{\text{SXoP}[r,n],u}^{\text{mu-prf}}(q_{\max}) &\leq \min\left(r^{r/2} \frac{\sqrt{uq_{\max}}}{N^{r/2-1/2}}, 2r^{r/2} \frac{uq_{\max}}{N^{r/2}}\right) \\ &\leq \min\left(O_r\left(\frac{\sqrt{uq_{\max}}}{N^{r/2-1/2}}\right), O_r\left(\frac{uq_{\max}}{N^{r/2}}\right)\right), \end{aligned} \quad (11)$$

where the first part of the second inequality (the first term inside min) also requires $r^{r/2} \frac{\sqrt{uq_{\max}}}{N^{r/2-1/2}} \leq \frac{1}{2}$.

Lower bound for even $r \geq 4$.

$$\text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) \geq 2^{-1} e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}} \geq \Omega_r\left(\frac{q}{N^{r/2}}\right). \quad (12)$$

Note that for $r \geq 4$, the theorem proves matching upper and lower single-user bounds of $\Theta_r\left(\frac{q}{N^{r/2}}\right)$. The bound for $r = 2$ and both bounds for odd $r \geq 3$ are known to be tight by previous works.

The proof relies on the following three lemmas regarding the density function $\nu_{n,k}^{(r)}$, generated by SXoP $[r, n]$.

Lemma 4 (\mathcal{L}^1 bidirectional bounds on $\hat{\nu}_{n,k}^{(r)}$ for even r). Assuming $k \leq N/4$ and r is even,

$$\frac{3k}{2\sqrt{r\binom{N}{r}}} \leq \mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \hat{\nu}_{n,k}^{(r)}(\alpha) \chi_{\alpha}(x) \right| \leq \frac{2k}{\sqrt{\binom{N}{r}}}.$$

Lemma 5 (Variance and weight bounds for $\nu_{n,q}^{(2)}$). Assume that $N \geq 100$ and $q \leq N/16$. Then,

$$\sum_{k=2}^q W^{=k}[\nu_{n,q}^{(2)}] \leq \frac{18q^2}{N^2}, \text{ and } \text{Var}[\nu_{n,q}^{(2)}] \leq \frac{4q}{N}.$$

Lemma 6 (Variance and weight bounds for $\nu_{n,q}^{(r)}$ with $r \geq 3$). Assume that $N \geq 2^{13}r$, $rq \leq N/8$. Then, for odd $r \geq 3$

$$\text{Var}[\nu_{n,q}^{(r)}] \leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-1}}.$$

For even $r \geq 4$,

$$\sum_{k=2}^q W^{=k}[\nu_{n,q}^{(r)}] \leq 2^{2r+1} r^{2r} \frac{q^2}{N^{2r-2}}, \text{ and } \text{Var}[\nu_{n,q}^{(r)}] \leq 2r^r \frac{q}{N^{r-1}}.$$

Proof (of Theorem 1). We prove the inequalities of the theorem.

Proof of (9). For r odd, by Lemma 6, $\text{Var}[\nu_{n,q}^{(r)}] \leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-1}}$. Both inequalities then follow by Proposition 7.

Proof of inequalities for even r . For even $r \geq 4$, we have $\text{Var}[\nu_{n,q}^{(r)}] \leq 2r^r \frac{q}{N^{r-1}}$ by Lemma 6. Combined with Proposition 7, this proves the first multi-user inequality of (11).

This variance bound gives a bound of $O_r\left(\frac{\sqrt{q}}{N^{(r-1)/2}}\right)$ on the statistical distance from uniform for $r \geq 4$, and a similar bound for $r = 2$ is obtained by Lemma 5. However, these bounds are not tight. For example, for $r = 2$, we obtain $O\left(\sqrt{\frac{q}{N}}\right)$, where the tight bound is known to be $O\left(\frac{q}{N}\right)$.

In order to improve the bound we use Proposition 5 with

$$\mathcal{S} = \mathcal{M}_{=1,q}^n = \{\alpha \in \widehat{\mathbb{F}}_2^{q \times n} \mid \#\alpha = 1\}.$$

Thus, combining (1) in Section 2 and Proposition 5 we obtain

$$\begin{aligned} 2\text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) &\leq 2\text{SD}(\nu_{n,q}^{(r)}, \mathbf{1}_{qn}) \\ &\leq \mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha) \chi_\alpha(x) \right| + \sqrt{\sum_{\alpha \in \mathcal{M}_{\geq 2,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha)^2}. \end{aligned} \quad (13)$$

By Lemma 4, the first term in (13) is bounded by

$$\mathbb{E}_{x \sim \mathbb{F}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha) \chi_\alpha(x) \right| \leq \frac{2q}{\sqrt{\binom{N}{r}}} \leq 2r^{r/2} \frac{q}{N^{r/2}}. \quad (14)$$

Proof of (10). For $r = 2$, by Lemma 5, the second term in (13) is bounded by

$$\sqrt{\sum_{\alpha \in \mathcal{M}_{\geq 2,q}^n} \widehat{\nu}_{n,q}^{(2)}(\alpha)^2} \leq \sqrt{\frac{18q^2}{N^2}} = \frac{\sqrt{18}q}{N}.$$

Therefore, using (13) with (14) and the bound on the second term above,

$$\text{Opt}_{\text{SXoP}[2,n]}^{\text{prf}}(q) \leq \frac{2q}{N} + \frac{\sqrt{4.5}q}{N} \leq \frac{5q}{N}.$$

Proof of (11). The first multi-user inequality of (11) was proved above. It remains to prove the single-user and second multi-user inequalities.

For $k \geq 4$, we apply Lemma 6 to bound the second term in (13) by

$$\begin{aligned} \sqrt{\sum_{\alpha \in \mathcal{M}_{\geq 2,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha)^2} &\leq \sqrt{2^{2r+1} r^{2r} \frac{q^2}{N^{2r-2}}} = 2^{r+1/2} r^r \frac{q}{N^{r-1}} = 2^{r+1/2} r^r \frac{1}{N^{r/2-1}} \frac{q}{N^{r/2}} \\ &\leq 2^{r+1/2} r^r \frac{1}{(2^{13r})^{r/2-1}} \frac{q}{N^{r/2}} = 2^{-5.5r+13.5} r^{r/2+1} \frac{q}{N^{r/2}}. \end{aligned} \quad (15)$$

where we have used the fact that $N \geq 2^{13}r$.

Therefore, using (13) with (14) and the bound on the second term above,

$$\begin{aligned} \text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) &\leq r^{r/2} \frac{q}{N^{r/2}} + 2^{-5.5r+12.5} r^{r/2+1} \frac{q}{N^{r/2}} \\ &= r^{r/2} \frac{q}{N^{r/2}} (1 + 2^{-5.5r+12.5} r) \leq 2r^{r/2} \frac{q}{N^{r/2}}, \end{aligned}$$

where we have used the fact that for $r \geq 4$, $2^{-5.5r+12.5} r \leq 1$. The second part of the multi-user bound of (11) (the second term inside min) follows from the single-user bound above by a straightforward triangle inequality.

Proof of (12). For the other direction, by Proposition 5,

$$2\text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) \geq \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha) \chi_{\alpha}(x) \right| - \sqrt{\sum_{\alpha \in \mathcal{M}_{\geq 2,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha)^2}.$$

By Lemma 4, the first term is lower bounded as

$$\left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\nu}_{n,q}^{(r)}(\alpha) \chi_{\alpha}(x) \right| \geq \frac{3q}{2\sqrt{r} \binom{N}{r}} \geq \frac{3}{2} e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}}.$$

Combining with the upper bound on the second term (15) we obtain

$$\begin{aligned} 2\text{Opt}_{\text{SXoP}[r,n]}^{\text{prf}}(q) &\geq \frac{3}{2} e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}} - 2^{-5.5r+13.5} r^{r/2+1} \frac{q}{N^{r/2}} \\ &\geq \frac{3}{2} e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}} (1 - 2^{-5.5r+13.5} e^{r/2} r^{3/2}) \\ &\geq \frac{3}{2} e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}} (1 - \frac{1}{3}) = e^{-r/2} r^{(r-1)/2} \frac{q}{N^{r/2}}, \end{aligned}$$

where the second inequality is based on the assumption $r \geq 4$. ■

Next, we prove the three lemmas.

4.1 Relation between $\widehat{\nu}_{n,k}^{(r)}$ and $\widehat{\mu}_{n,rk}$

We start by proving an elementary result that establishes the connection between the Fourier coefficients of $\nu_{n,k}^{(r)}$ and those of $\mu_{n,rk}$.

Proposition 17 (Relation between $\widehat{\nu}_{n,k}^{(r)}$ and $\widehat{\mu}_{n,rk}$). *For any $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$,*

$$\widehat{\nu}_{n,k}^{(r)}(\alpha) = \widehat{\mu}_{n,rk}(\alpha^{\circ r}) = \widehat{\mu}_{n,rk}(\alpha^{\odot r}).$$

Proof. By definition of SXoP $[r, n]$ and Proposition 2, for any $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$

$$\begin{aligned} \widehat{\nu}_{n,k}^{(r)}(\alpha) &= \mathbb{E}_{x \sim \nu_{n,k}^{(r)}} [\chi_{\alpha}(x)] = \mathbb{E}_{y_{1..k}^1 \dots y_{1..k}^r \sim \mu_{n,rk}} [\chi_{\alpha}(\oplus_{\ell=1}^r y_{1..k}^{\ell})] \\ &= \mathbb{E}_{y_{1..k}^1 \dots y_{1..k}^r \sim \mu_{n,rk}} [\chi_{\alpha^{\circ r}}(y_{1..k}^1, \dots, y_{1..k}^r)] = \widehat{\mu}_{n,rk}(\alpha^{\circ r}). \end{aligned}$$

Finally, $\widehat{\mu}_{n,rk}(\alpha^{\circ r}) = \widehat{\mu}_{n,rk}(\alpha^{\odot r})$ holds by Proposition 8. ■

4.2 Proof of Lemma 4

Proof (of Lemma 4). Let $\beta \in \widehat{\mathbb{F}}_2^{k \times n}$ be any fixed mask with $\#\beta = 1$. Also, let $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$ be a mask with $\#\alpha = 1$. Observe that $\alpha^{\circ r}$ has $\#(\alpha^{\circ r}) = r$ and all its r non-zero elements are equal. Since r is even, by Proposition 12, $\widehat{\mu}_{n,rk}(\alpha^{\circ r}) = \widehat{\mu}_{n,rk}(\beta^{\circ r})$ is independent of the actual non-zero element. Therefore, applying Proposition 17 and Proposition 12,

$$\begin{aligned} & \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \widehat{\nu}_{n,k}^{(r)}(\alpha) \chi_\alpha(x) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\circ r}) \chi_\alpha(x) \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \widehat{\mu}_{n,rk}(\beta^{\circ r}) \chi_\alpha(x) \right| = |\widehat{\mu}_{n,rk}(\beta^{\circ r})| \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right|. \end{aligned}$$

By Proposition 12, $\frac{1}{\sqrt{r \binom{N}{r}}} \leq |\widehat{\mu}_{n,rk}(\beta^{\circ r})| \leq \frac{1}{\sqrt{\binom{N}{r}}}$.

It thus remains to prove that

$$\frac{3k}{2} \leq \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right| \leq 2k. \quad (16)$$

Note that the expression $\mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right|$ is not directly related to the density functions $\nu_{n,k}^{(r)}$ and $\mu_{n,rk}$.

For every $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$ with $\#\alpha = 1$, define $in(\alpha)$ to be the unique index i with $\alpha_i \neq 0$. Then,

$$\begin{aligned} & \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \prod_{i \in [k]} \chi_{\alpha_i}(x_i) \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_{\alpha_{in(\alpha)}}(x_{in(\alpha)}) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{i=1}^k \sum_{\substack{\gamma \in \widehat{\mathbb{F}}_2^n \\ \gamma \neq 0}} \chi_\gamma(x_i) \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{i=1}^k \left(\sum_{\gamma \in \widehat{\mathbb{F}}_2^n} \chi_\gamma(x_i) - \chi_0(x_i) \right) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| N \sum_{i=1}^k \left(\mathbb{E}_{\gamma \in \widehat{\mathbb{F}}_2^n} [\chi_\gamma(x_i)] \right) - k \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| N \sum_{i=1}^k (\mathbb{1}(x_i = 0)) - k \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} |N|\{i \in [k] \mid x_i = 0\}| - k|, \end{aligned}$$

where the penultimate equality is by orthogonality of the characters.

Therefore, denoting $Z_x = |\{i \in [k] \mid x_i = 0\}|$,

$$\mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} |N \cdot Z_x - k| = N \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{k \times n}} [|Z_x - \frac{k}{N}|].$$

Observe that the random variable Z_x follows a binomial distribution with number of experiments k and success probability $\frac{1}{N}$, and thus satisfies $\mathbb{E}[Z_x] = \frac{k}{N}$.

Therefore, $\mathbb{E}_x[|Z_x - \frac{k}{N}|] = \mathbb{E}_x[|Z_x - \mathbb{E}[Z_x]|]$. So far, we have obtained

$$\mathbb{E}_{x \sim \mathbb{F}_2^{k \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,k}^n} \chi_\alpha(x) \right| = N \mathbb{E}_x[|Z_x - \mathbb{E}[Z_x]|],$$

and by (16) it remains to prove that $\frac{3k}{2N} \leq \mathbb{E}_x[|Z_x - \mathbb{E}[Z_x]|] \leq \frac{2k}{N}$.

We have

$$\begin{aligned} \mathbb{E}_x[|Z_x - \mathbb{E}[Z_x]|] &= \Pr[Z_x = 0] \frac{k}{N} + \sum_{s=1}^k \Pr[Z_x = s] (s - \frac{k}{N}) \\ &= 2 \Pr[Z_x = 0] \frac{k}{N} + \sum_{s=0}^k \Pr[Z_x = s] (s - \frac{k}{N}) = 2 \Pr[Z_x = 0] \frac{k}{N} + \mathbb{E}[Z_x - \mathbb{E}[Z_x]] \\ &= 2 \frac{k}{N} \Pr[Z_x = 0] + 0 = 2 \frac{k}{N} (1 - \frac{1}{N})^k. \end{aligned}$$

Finally, $(1 - \frac{1}{N})^k \leq 1$ and as $k \leq \frac{N}{4}$, $(1 - \frac{1}{N})^k \geq (1 - \frac{1}{N})^{N/4} \geq \frac{3}{4}$. \blacksquare

4.3 Basic Results and Proof of Lemma 5

We prove simple bounds that are similar to (3) and (4), proved in the motivating example of Section 3. We then use these results to prove Lemma 5.

Proposition 18 (Bound 1 on level- k Fourier weight of $\nu_{n,q}^{(r)}$). *Assume that $rq \leq N/2$. Then, for even r*

$$\mathbb{W}^{=k}[\nu_{n,q}^{(r)}] \leq \binom{q}{k} N^k \frac{1}{\binom{N}{rk}} \leq \binom{q}{k} (rk)^k \left(\frac{rk}{N}\right)^{(r-1)k}.$$

For odd r , $\mathbb{W}^{=1}[\nu_{n,q}^{(r)}] = 0$ and

$$\mathbb{W}^{=k}[\nu_{n,q}^{(r)}] \leq \binom{q}{k} N^{k-1} \frac{1}{\binom{N}{rk}} \leq \binom{q}{k} (rk)^{k-1} \left(\frac{rk}{N}\right)^{(r-1)k+1}.$$

Proof. Applying Proposition 17 and then Lemma 1,

$$\begin{aligned} \mathbb{W}^{=k}[\nu_{n,q}^{(r)}] &= \binom{q}{k} \mathbb{W}^{=k}[\nu_{n,k}^{(r)}] = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\nu}_{n,k}^{(r)}(\alpha)^2 \\ &= \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 \leq \binom{q}{k} N^k M^{=rk}[\mu_{n,rk}] \leq \binom{q}{k} N^k \frac{1}{\binom{N}{rk}}. \end{aligned} \quad (17)$$

When r is odd, then by Proposition 9, $\alpha^{\odot r} \neq 0$ only if $0 = \oplus_{i \in [rk]} (\alpha^{\odot r})_i = \oplus_{i \in [k]} \alpha_i$, which holds only for at most N^{k-1} of the masks in $\mathcal{M}_{=k,k}^n \subset \widehat{\mathbb{F}}_2^{k \times n}$. Hence for odd r the bound is improved by a factor of N .

For the particular case of $k = 1$, we have $\oplus_{i \in [k]} \alpha_i \neq 0$ when $\#\alpha = 1$, and hence $\mathbb{W}^{=1}[\nu_{n,q}^{(r)}] = 0$. \blacksquare

Proposition 19 (Bound 2 on level- k Fourier weight of $\nu_{n,q}^{(r)}$). Assume that $rq \leq N/2$. Then,

$$W^{=k}[\nu_{n,q}^{(r)}] \leq \binom{q}{k} \left(\frac{rk}{N-rk} \right)^{rk/2} \leq \binom{q}{k} \left(\frac{2rk}{N} \right)^{rk/2}.$$

We remark that Proposition 18 gives a better bound than Proposition 19 for small values of k , while Proposition 19 is better for large values of k . However, both are very far from being tight in general.

Proof. Applying Proposition 17 (similarly to (17) above) and then Lemma 2,

$$\begin{aligned} W^{=k}[\nu_{n,q}^{(r)}] &= \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2 \leq \binom{q}{k} \sum_{\beta \in \mathcal{M}_{=rk,rk}^n} \widehat{\mu}_{n,rk}(\beta)^2 \\ &= \binom{q}{k} W^{=rk}[\mu_{n,rk}] \leq \binom{q}{k} \left(\frac{rk}{N-rk} \right)^{rk/2}. \end{aligned}$$

■

Proof (of Lemma 5). By Proposition 19, for $r = 2$,

$$\sum_{k=2}^q W^{=k}[\nu_{n,q}^{(2)}] \leq \sum_{k=2}^q \binom{q}{k} \left(\frac{2k}{N-2k} \right)^k = \sum_{k=2}^q \binom{q}{k} 2^k \left(\frac{k}{N-2k} \right)^k.$$

Denote $B(k) = \binom{q}{k} 2^k \left(\frac{k}{N-2k} \right)^k$. Assuming $k+1 \leq q$, by Proposition 1,

$$\frac{B(k+1)}{B(k)} \leq 2 \frac{q}{k+1} \frac{k+1}{N-2k-2} e^{\frac{k}{k} + \frac{2k}{N-2k-2}} \leq \frac{2q}{N-2q} e^{1 + \frac{2q}{N-2q}} \leq \frac{1}{7} e^{8/7} \leq \frac{1}{2},$$

as $q \leq N/16$. Therefore,

$$\sum_{k=2}^q W^{=k}[\nu_{n,q}^{(r)}] \leq \sum_{k=2}^q B(k) \leq 2B(2) = 2 \binom{q}{2} 2^2 \frac{2^2}{(N-4)^2} \leq 18 \left(\frac{q}{N} \right)^2,$$

as $N \geq 100$. Combining with Proposition 18 that asserts $W^{=1}[\nu_{n,q}^{(2)}] \leq qN \frac{1}{\binom{N}{2}} \leq 2 \frac{q}{N}$, we deduce

$$\text{Var}[\nu_{n,q}^{(2)}] = \sum_{k=1}^q W^{=k}[\nu_{n,q}^{(2)}] \leq 2 \frac{q}{N} + 18 \frac{q^2}{N^2} \leq 2 \frac{q}{N} + \frac{18}{16} \frac{q}{N} \leq \frac{4q}{N},$$

as $\frac{q}{N} \leq \frac{1}{16}$. ■

4.4 Application of Main Framework and Proof of Lemma 6

We apply our main framework and use it to prove Lemma 6.

Proposition 20. Assume that $rq \leq N/8$ and $k \geq 2$. Define $c_{rk} = 0$ if rk is even and $c_{rk} = \frac{1}{2}$ if rk is odd (i.e., $c_{rk} = \frac{rk \bmod 2}{2}$). Then, for any $r \geq 3$

$$W^{=k}[\nu_{n,q}^{(r)}] \leq \binom{q}{k} 2^{(r-2)k/2 + c_{rk}} \left(\frac{rk}{N-rk} \right)^{(r-1)k + c_{rk}}.$$

Proof. Applying Proposition 17,

$$W^{=k}[\nu_{n,q}^{(r)}] = \binom{q}{k} W^{=k}[\nu_{n,k}^{(r)}] = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\nu}_{n,k}^{(r)}(\alpha)^2 = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2. \quad (18)$$

We would like to bound $\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,rk}(\alpha^{\odot r})^2$ using Lemma 3. We start by introducing several definitions referring to Setting 1 and then define the PIS, *pis*. First, define, $\mathcal{S} = \mathcal{M}_{=k,k}^n = \{\alpha \in \mathbb{F}_2^{k \times n} \mid \#\alpha = k\}$, and $T(\alpha) = T_{r,k}(\alpha) = \alpha^{\odot r} = ((\alpha_1)^{\odot r}, \dots, (\alpha_k)^{\odot r})$ (here $k_0 = k' = rk$).

Given an index $\ell \in [k]$, for all $\ell' \in [(\ell-1)r+1, \ell r]$, $T(\alpha)_{\ell'} = (\alpha^{\odot r})_{\ell'} = \alpha_{\ell}$. Thus, for a recursion node v , if $\ell' \in [(\ell-1)r+1, \ell r]$ is unaltered by Definition 7, then $T_v(\alpha)_{\ell'} = T(\alpha)_{\ell'} = \alpha_{\ell}$ for every $\alpha \in \mathcal{S}_v$. We call an index $\ell \in [k]$ redundant (for a node v) if at least 3 of the r indices in $[(\ell-1)r+1, \ell r]$ are unaltered.

Given a recursion node v , let $\ell \in [k]$ be the largest redundant index. The PIS *pis* selects as primary index the largest unaltered index $j \in [(\ell-1)r+1, \ell r]$.

Let $d = d_{\max} = \lceil \frac{(r-2)k}{2} \rceil$. We first prove that there is always a redundant index for nodes up to depth $d-1 = \lceil \frac{(r-2)k}{2} \rceil - 1$.

The number of unaltered indices at depth $d-1$ is at least $k' - 2(d-1) = rk - 2\lceil \frac{(r-2)k}{2} \rceil + 2 \geq rk - ((r-2)k+1) + 2 = 2k+1$. By an averaging argument, there exists $\ell \in [k]$ such that $[(\ell-1)r+1, \ell r]$ contains at least $\lceil \frac{2k+1}{k} \rceil = 3$ unaltered indices. Namely, ℓ is redundant. This proves that *pis* is well-defined up to depth $d = \lceil \frac{(r-2)k}{2} \rceil$ (at the leaves of depth d we do not invoke *pis*).

In order to apply Lemma 3, it is sufficient to prove that the two conditions of Proposition 13 hold. Clearly, the pair (\mathcal{S}, T) satisfies the restrictions of Setting 1, and condition (c1) holds.

We now prove condition (c2). Specifically, we prove that for a node v such that $j = \text{pis}(v)$ and $\alpha \in \mathcal{S}_v$, $T_v(\alpha)_j$ can be computed from $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ (where i is a secondary index).

For a node v we select a primary index $j \in [rk]$ such that $T_v(\alpha)_j = \alpha_{\ell}$ and since ℓ is redundant, $T_v(\alpha)_{\ell'} = \alpha_{\ell}$ for at least 3 indices $\ell' \in [(\ell-1)r+1, \ell r]$. As $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$, and $T_v(\alpha)^{\oplus(j,i)}$ modifies 2 entries of $T_v(\alpha)$, then $\alpha_{\ell} = T_v(\alpha)_j$ still appears at least $3-2=1$ time in $T_{v_i}(\alpha)$. This proves condition (c2) as required.

Applying our framework of Lemma 3 (with $d = \lceil \frac{(r-2)k}{2} \rceil = \frac{(r-2)k}{2} + c_{rk}$, $k_0 = k' = rk$), we obtain

$$\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,r}(\alpha^{\odot r})^2 \leq 2^d \left(\frac{k'}{N-k'}\right)^{k'/2+d} = 2^{(r-2)k/2+c_{rk}} \left(\frac{rk}{N-rk}\right)^{(r-1)k+c_{rk}}.$$

Combining with (18) completes the proof. \blacksquare

The proof of Lemma 6 uses the bounds on $W^{=k}[\nu_{n,q}^{(r)}]$ of Proposition 20 and additional simple bounds to analyze several sums of weights. It essentially shows that in all cases the lowest-level weight bound dominates the sum (except for r odd, as for $k=1$, $W^{=1}[\nu_{n,q}^{(r)}] = 0$).

Proof (of Lemma 6). Consider any $r \geq 3$. By Proposition 20,

$$W^{=k}[\nu_{n,q}^{(r)}] \leq \binom{q}{k} 2^{(r-2)k/2+c_{rk}} \left(\frac{rk}{N-rk}\right)^{(r-1)k+c_{rk}}.$$

Write $N = Mr$ and define

$$B_r(k) = \binom{q}{k} 2^{(r-2)k/2+c_{rk}} \left(\frac{rk}{N-rk}\right)^{(r-1)k+c_{rk}} = \binom{q}{k} 2^{(r-2)k/2+c_{rk}} \left(\frac{k}{M-k}\right)^{(r-1)k+c_{rk}}.$$

Then, for $2 \leq k \leq q-2$ (noting that $c_{rk} = c_{r(k+2)}$), by Proposition 1,

$$\begin{aligned} \frac{B_r(k+2)}{B_r(k)} &\leq \frac{q^2}{(k+1)(k+2)} 2^{r-2} e^{\frac{2((r-1)k+c_{rk})}{k} + \frac{2((r-1)k+c_{rk})}{M-k-2}} \left(\frac{k+2}{M-k-2}\right)^{2(r-1)} \\ &\leq 2^{r-2} e^{2(r-1) + \frac{1}{k} + \frac{2(r-1)q}{M-q} \frac{k+2}{k+1} \frac{q^2}{(k+2)^2} \left(\frac{k+2}{M-q}\right)^{2(r-1)}} \\ &\leq 2^{r-2} \frac{k+2}{k+1} e^{\frac{16(r-1)}{7} + \frac{1}{k}} \left(\frac{q}{7q}\right)^{2(r-1)} \leq \frac{4}{3} \left(\frac{2e^{16/7}}{49}\right)^{r-1} \frac{1}{2} e^{\frac{1}{2}} \leq \left(\frac{1}{2}\right)^2 \frac{2}{3} e^{\frac{1}{2}} \leq \frac{1}{2}, \end{aligned}$$

where we have used the facts $k \geq 2$, $M = \frac{N}{r} \geq 8q$ and $r \geq 3$.

Therefore, using the fact that $N \geq 2^{13}r$,

$$\begin{aligned} \sum_{k=3}^q W^{=k}[\nu_{n,q}^{(r)}] &\leq \sum_{k=3}^q B_r(k) \leq 2B_r(3) + 2B_r(4) \\ &= 2\binom{q}{3} 2^{3(r-2)/2+c_{3r}} \left(\frac{3r}{N-3r}\right)^{3(r-1)+c_{3r}} + 2\binom{q}{4} 2^{2(r-2)} \left(\frac{4r}{N-4r}\right)^{4(r-1)} \\ &\leq 2^{3r/2-4+c_{3r}} q^3 \left(\frac{4r}{N}\right)^{3(r-1)+c_{3r}} + 2^{2r-7} q^4 \left(\frac{8r}{N}\right)^{4(r-1)} \\ &= 2^{7.5r-10+3c_{3r}} r^{3r-3+c_{3r}} \frac{q^3}{N^{3r-3+c_{3r}}} + 2^{14r-19} r^{4r-4} \frac{q^4}{N^{4r-4}} \\ &= 2^{7.5r-10+3c_{3r}} r^{3r-3+c_{3r}} \frac{q^3}{N^{3r-3+c_{3r}}} \left(1 + 2^{6.5r-9-3c_{3r}} r^{r-1-c_{3r}} \frac{q}{N^{r-1-c_{3r}}}\right). \end{aligned}$$

We have

$$\begin{aligned} 2^{6.5r-9-3c_{3r}} r^{r-1-c_{3r}} \frac{q}{N^{r-1-c_{3r}}} &= 2^{6.5r-9-3c_{3r}} \left(\frac{r}{N}\right)^{r-2-c_{3r}} \frac{rq}{N} \\ &\leq 2^{6.5r-9-3c_{3r}} 2^{-13(r-2-c_{3r})} \frac{1}{8} = 2^{-6.5r-9+26-3+10c_r} \\ &\leq 2^{-6.5r+14+10c_r} \leq 2^{-6.5r+19} \leq 1, \end{aligned}$$

where we have used the assumptions $r \geq 3$, $rq \leq N/8$ and $N \geq 2^{13}r$. Plugging this into the previous inequality, we deduce

$$\sum_{k=3}^q W^{=k}[\nu_{n,q}^{(r)}] \leq 2^{7.5r-9+3c_{3r}} r^{3r-3+c_{3r}} \frac{q^3}{N^{3r-3+c_{3r}}}. \quad (19)$$

Assume that r is odd. Then, by Proposition 18, $W^{=1}[\nu_{n,q}^{(r)}] = 0$. Moreover, by Proposition 18 (which gives a better bound on $W^{=2}[\nu_{n,q}^{(r)}]$ than Proposition 20)

$$W^{=2}[\nu_{n,q}^{(r)}] \leq \binom{q}{2} (2r) \left(\frac{2r}{N}\right)^{2r-1} \leq 2^{2r-1} r^{2r} \frac{q^2}{N^{2r-1}}.$$

Hence by the above results and (19) (noting that $c_{3r} = \frac{1}{2}$ and recalling that $rq \leq N/8$),

$$\begin{aligned}\text{Var}[\nu_{n,q}^{(r)}] &= \sum_{k=2}^q W^{=k}[\nu_{n,q}^{(r)}] \leq 2^{2r-1} r^{2r} \frac{q^2}{N^{2r-1}} + 2^{7.5r-9+1.5} r^{3r-3+0.5} \frac{q^3}{N^{3r-3+0.5}} \\ &= 2^{2r-1} r^{2r} \frac{q^2}{N^{2r-1}} \left(1 + 2^{5.5r-6.5} r^{r-2.5} \frac{q}{N^{r-1.5}}\right).\end{aligned}$$

We have

$$\begin{aligned}2^{5.5r-6.5} r^{r-2.5} \frac{q}{N^{r-1.5}} &= 2^{5.5r-6.5} \left(\frac{r}{N}\right)^{r-2.5} \frac{q}{N} \leq 2^{5.5r-6.5} 2^{-13(r-2.5)} \frac{1}{8.3} \\ &\leq 2^{-7.5r-6.5+32.5-4} = 2^{-7.5r+22} \leq 1,\end{aligned}$$

where we have used the assumptions $r \geq 3$, $rq \leq N/8$ and $N \geq 2^{13}r$. Plugging this into the previous inequality, we deduce the claimed inequality

$$\text{Var}[\nu_{n,q}^{(r)}] \leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-1}}.$$

For even $r \geq 4$, by Proposition 18, $W^{=2}[\nu_{n,q}^{(r)}] \leq \binom{q}{2} (2r)^2 \left(\frac{2r}{N}\right)^{2r-2} = 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}}$. Therefore, by the above inequality and (19) (with $c_{rk} = 0$),

$$\begin{aligned}\sum_{k=2}^q W^{=k}[\nu_{n,q}^{(r)}] &\leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} + 2^{7.5r-9} r^{3r-3} \frac{q^3}{N^{3r-3}} \\ &= 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} \left(1 + 2^{5.5r-9} r^{r-3} \frac{q}{N^{r-1}}\right) = 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} \left(1 + 2^{5.5r-9} \left(\frac{r}{N}\right)^{r-3} \frac{q}{N} \frac{1}{N}\right) \\ &\leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} \left(1 + 2^{5.5r-9} 2^{-13(r-3)} \frac{1}{8.4} 2^{-13}\right) \leq 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} \left(1 + 2^{-7.5r-9+39-5-13}\right) \\ &= 2^{2r} r^{2r} \frac{q^2}{N^{2r-2}} \left(1 + 2^{-7.5r+12}\right) \leq 2^{2r+1} r^{2r} \frac{q^2}{N^{2r-2}},\end{aligned}$$

where we have used the assumptions $r \geq 4$, $rq \leq N/8$ and $N \geq 2^{13}r$.

Finally, by Proposition 18 and the above inequality (again using the assumptions $r \geq 4$, $rq \leq N/8$ and $N \geq 2^{13}r$),

$$\begin{aligned}\text{Var}[\nu_{n,q}^{(r)}] &= \sum_{k=1}^q W^{=k}[\nu_{n,q}^{(r)}] \leq q r^r \frac{1}{N^{r-1}} + 2^{2r+1} r^{2r} \frac{q^2}{N^{2r-2}} \\ &= r^r \frac{q}{N^{r-1}} \left(1 + 2^{2r+1} r^r \frac{q}{N^{r-1}}\right) \leq r^r \frac{q}{N^{r-1}} \left(1 + 2^{2r+1} r \frac{rq}{N} \left(\frac{r}{N}\right)^{r-2}\right) \\ &\leq r^r \frac{q}{N^{r-1}} \left(1 + 2^{2r+1} r \frac{1}{8} 2^{-13(r-2)}\right) = r^r \frac{q}{N^{r-1}} \left(1 + 2^{-11r+1-3+26} r\right) \\ &= r^r \frac{q}{N^{r-1}} \left(1 + 2^{-11r+24} r\right) \leq 2r^r \frac{q}{N^{r-1}}.\end{aligned}$$

■

5 Indistinguishability Bounds for LXoP[L, n]

In this section, we state and prove our main theorem regarding LXoP[L, n].

Theorem 2. Assume that the function $L'(x) = x \oplus L(x)$ is a permutation on \mathbb{F}_2^n . Given that $N \geq 2^{10}$ and $q \leq N/16$,

$$\text{Opt}_{\text{LXoP}[L,n]}^{\text{prf}}(q) \leq \frac{4q}{N^{1.5}},$$

and assuming $\frac{6\sqrt{u}q_{\max}}{N^{1.5}} \leq \frac{1}{2}$,

$$\text{Opt}_{\text{LXoP}[L,n],u}^{\text{mu-prf}}(q_{\max}) \leq \frac{6\sqrt{u}q_{\max}}{N^{1.5}}.$$

The proof is based on the following lemma, proved in the remainder of this section.

Lemma 7. Assume that the function $L'(x) = x \oplus L(x)$ is a permutation on \mathbb{F}_2^n . Given that $N \geq 2^{10}$ and $q \leq N/16$, $\text{Var}[\xi_{n,q}^{(L)}] \leq \frac{64q^2}{N^3}$.

Proof (of Theorem 2). The proof is immediate from Lemma 7 and Proposition 7. \blacksquare

5.1 Elementary results

We first establish the connection between the Fourier coefficients of $\xi_{n,k}^{(L)}$ and those of $\mu_{n,2k}$.

Proposition 21 (Relation between $\widehat{\xi}_{n,k}^{(L)}$ and $\widehat{\mu}_{n,2k}$). For any $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$,

$$\widehat{\xi}_{n,k}^{(L)}(\alpha) = \widehat{\mu}_{n,2k}(\alpha, L^{\text{T}}(\alpha)) = \widehat{\mu}_{n,2k}(\alpha_1, L^{\text{T}}(\alpha_1), \dots, \alpha_k, L^{\text{T}}(\alpha_k)).$$

Proof. By definition of $\text{LXoP}[L, n]$ and Proposition 2, for any $\alpha \in \widehat{\mathbb{F}}_2^{k \times n}$

$$\begin{aligned} \widehat{\xi}_{n,k}^{(L)}(\alpha) &= \mathbb{E}_{x \sim \xi_{n,k}^{(L)}} [\chi_{\alpha}(x)] = \mathbb{E}_{y_{1..k}^{1,2} \sim \mu_{n,2k}} [\chi_{\alpha}(y_1^1 \oplus L(y_1^2), \dots, y_k^1 \oplus L(y_k^2))] \\ &= \mathbb{E}_{y_{1..k}^{1,2} \sim \mu_{n,2k}} [\chi_{\alpha, \alpha}(y_{1..k}^1, L(y_{1..k}^2))] \\ &= \mathbb{E}_{y_{1..k}^{1,2} \sim \mu_{n,2k}} [\chi_{\alpha, L^{\text{T}}(\alpha)}(y_{1..k}^1, y_{1..k}^2)] = \widehat{\mu}_{n,2k}(\alpha, L^{\text{T}}(\alpha)). \end{aligned}$$

Proposition 22. Assume that the function $L'(x) = x \oplus L(x)$ is a permutation on \mathbb{F}_2^n . Then, $W^1[\xi_{n,q}^{(L)}] = 0$.

Proof. By Proposition 21,

$$W^1[\xi_{n,q}^{(L)}] = \binom{q}{1} \sum_{\alpha \in \mathcal{M}_{=1,1}^n} \widehat{\xi}_{n,k}^{(L)}(\alpha)^2 = q \sum_{\alpha \in \mathcal{M}_{=1,1}^n} \widehat{\mu}_{n,2}(\alpha, L^{\text{T}}(\alpha))^2.$$

Let $\alpha \in \widehat{\mathbb{F}}_2^n$ be non-zero. We have $\alpha \oplus L^{\text{T}}(\alpha) = (L')^{\text{T}}(\alpha)$. Since L is a permutation, so is $(L')^{\text{T}}$. Since $(L')^{\text{T}}(0) = 0$, this implies that $(L')^{\text{T}}(\alpha) \neq 0$, hence $\alpha \oplus L^{\text{T}}(\alpha) \neq 0$. By proposition 9, we deduce $\widehat{\mu}_{n,2}(\alpha, L^{\text{T}}(\alpha)) = 0$, implying $W^1[\xi_{n,q}^{(L)}] = 0$. \blacksquare

5.2 Application of Main Framework and Proof of Lemma 7

We use our main framework to prove Lemma 7.

Proposition 23. *Assume that the function $L'(x) = x \oplus L(x)$ is a permutation on \mathbb{F}_2^n , and assume that $k \leq N/16$. Define $c_k = 0$ if k is even and $c_k = \frac{1}{2}$ otherwise (i.e. $c_k = \frac{k \bmod 2}{2}$). Then,*

$$W^{=k}[\xi_{n,q}^{(L)}] \leq \binom{q}{k} 2^{3k/2+3c_k} \frac{k^{k+2c_k} (k-2c_k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}}.$$

Proof. Based on Proposition 21,

$$\begin{aligned} W^{=k}[\xi_{n,q}^{(L)}] &= \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\xi}_{n,k}^{(L)}(\alpha)^2 = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,2k}(\alpha, L^T(\alpha))^2 \\ &= \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,2k}(\alpha_1, L^T(\alpha_1), \dots, \alpha_k, L^T(\alpha_k))^2. \end{aligned} \quad (20)$$

We now upper bound $\sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,2k}(\alpha_1, L^T(\alpha_1), \dots, \alpha_k, L^T(\alpha_k))^2$ using Lemma 3.

We start by introducing some definitions referring to Setting 1 and then define the PIS *pis*. Define, $\mathcal{S} = \mathcal{M}_{=k,k}^n = \{\alpha \in \widehat{\mathbb{F}}_2^{k \times n} \mid \#\alpha = k\}$, and $T(\alpha) = T_k(\alpha) = (\alpha_1, L^T(\alpha_1), \dots, \alpha_k, L^T(\alpha_k))$ (here $k_0 = k' = 2k$).

Given a node v , we say that an index $\ell \in [k]$ is redundant if both $2\ell - 1$ and 2ℓ are unaltered by Definition 7. Note that if ℓ is redundant, then for every $\alpha \in \mathcal{S}_v$, $T_v(\alpha)_{2\ell-1} = \alpha_\ell$ and $T_v(\alpha)_{2\ell} = L^T(\alpha_\ell)$.

At a given recursion node v , the PIS *pis* will select as primary index the largest index $2\ell - 1$ such that $\ell \in [k]$ is redundant.

Let $d = d_{\max} = \lceil k/2 \rceil$. We first prove that there is always a redundant index for nodes up to depth $d - 1 = \lceil k/2 \rceil - 1$. Indeed, every recursive call can remove at most 2 redundant indices, and thus at depth $d - 1$, we have at least $k - 2(d - 1) = k - 2\lceil k/2 \rceil + 2 \geq 1$ redundant indices. This proves that *pis* is well-defined up to depth $d = \lceil k/2 \rceil$ (at the leaves of depth d we do not invoke *pis*).

In order to apply Lemma 3, we prove that the two conditions of Proposition 13 hold. First, the pair (\mathcal{S}, T) satisfies the restrictions of Setting 1, and condition (c1) holds.

We now prove condition (c2). Namely, for a node v such that $j = \text{pis}(v)$ and $\alpha \in \mathcal{S}_v$, we prove that $T_v(\alpha)_j$ can be computed from $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ (where i is a secondary index).

For a node v we select as primary index $j = 2\ell - 1$ for $\ell \in [k]$ redundant, and we have $T_v(\alpha)_j = T_v(\alpha)_{2\ell-1} = \alpha_\ell$ and $T_v(\alpha)_{2\ell} = L^T(\alpha_\ell)$.

If $i \neq 2\ell$, then

$$T_{v_i}(\alpha)_{2\ell} = T_v(\alpha)_{2\ell} = L^T(\alpha_\ell),$$

and we can compute $L^{-T}(T_{v_i}(\alpha)_{2\ell}) = \alpha_\ell = T_v(\alpha)_j$. Otherwise, $i = 2\ell$, and

$$T_{v_i}(\alpha)_{2\ell} = T_v(\alpha)_{2\ell-1} \oplus T_v(\alpha)_{2\ell} = \alpha_\ell \oplus L^T(\alpha_\ell) = (L')^T(\alpha_\ell).$$

Since $(L')^T$ is an invertible linear transformation, we can compute $(L')^{-T}(T_{v_i}(\alpha)_{2\ell}) = \alpha_\ell = T_v(\alpha)_j$. This proves condition **(c2)**.

Applying our framework of Lemma 3 (with $d = \lceil k/2 \rceil = k/2 + c_k$, $k_0 = k' = 2k$), we obtain

$$\begin{aligned} \sum_{\alpha \in \mathcal{M}_{=k,k}^n} \widehat{\mu}_{n,2k}(\alpha_1, L^T(\alpha_1), \dots, \alpha_k, L^T(\alpha_k))^2 &\leq 2^d \frac{(k')^{2d} (k'-2d)^{k'/2-d}}{(N-k')^{k'/2+d}} \\ &= 2^{k/2+c_k} \frac{(2k)^{k+2c_k} (k-2c_k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}} = 2^{3k/2+3c_k} \frac{k^{k+2c_k} (k-2c_k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}}. \end{aligned}$$

Combining with (20) completes the proof. \blacksquare

The proof of Lemma 7 uses Proposition 23 and shows that the bound on $W^{=2}[\xi_{n,q}^{(L)}]$ dominates $\text{Var}[\xi_{n,q}^{(L)}]$.

Proof (of Lemma 7). By Proposition 22, $W^{-1}[\xi_{n,q}^{(L)}] = 0$. Hence, by Proposition 23,

$$\text{Var}[\xi_{n,q}^{(L)}] = \sum_{k=1}^n W^{=k}[\xi_{n,q}^{(L)}] = \sum_{k=2}^n W^{=k}[\xi_{n,q}^{(L)}] \leq \sum_{k=2}^n \binom{q}{k} 2^{3k/2+3c_k} \frac{k^{k+2c_k} (k-2c_k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}}.$$

Denote $B(k) = \binom{q}{k} 2^{3k/2+3c_k} \frac{k^{k+2c_k} (k-2c_k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}}$. Noting that $c_k = c_{k+2}$ and assuming $2 \leq k \leq q-2$, by Proposition 1,

$$\begin{aligned} \frac{B(k+2)}{B(k)} &\leq \frac{q^2}{(k+1)(k+2)} 2^3 e^{\frac{2(k+2c_k)}{k} + \frac{2(k/2-c_k)}{k-2c_k} + \frac{2^2(3k/2+c_k)}{N-2k-4} \frac{(k+2)^2(k+2-2c_k)}{(N-2k-4)^3}} \\ &\leq 8q^2 e^{2+\frac{4c_k}{k}+1+\frac{6k+2}{N-2k-4} \frac{k+2}{k+1} \frac{k+2-2c_k}{(N-2k-4)^3}} \leq 8\frac{4}{3} e^{3+\frac{2}{3}+\frac{6q}{14q}} q^2 \frac{q}{(14q)^3} \leq 8\frac{4}{3} e^{4.1} 14^{-3} \leq \frac{1}{3}, \end{aligned}$$

where we have used the facts $k \geq 2$, $\frac{4c_k}{k} \leq \frac{2}{3}$ and $q \leq N/16$. Therefore,

$$\begin{aligned} \sum_{k=2}^n W^{=k}[\xi_{n,q}^{(L)}] &\leq \frac{1}{1-\frac{1}{3}} (B(2) + B(3)) \leq \frac{3}{2} \binom{q}{2} 2^3 \frac{2^3}{(N-4)^3} + \frac{3}{2} \binom{q}{3} 2^6 \frac{3^4 \cdot 2}{(N-6)^5} \\ &\leq \frac{48q^2}{(N-4)^3} + \frac{2^5 3^4 q^3}{(N-6)^5} \leq 60 \frac{q^2}{N^3} + 2^6 3^4 \frac{q}{N} \frac{1}{N} \frac{q^2}{N^3} \leq 60 \frac{q^2}{N^3} + \frac{q^2}{N^3} \leq 2^6 \frac{q^2}{N^3}, \end{aligned}$$

where we have used the assumptions that $q \leq N/16$ and $N \geq 2^{10}$. \blacksquare

6 Indistinguishability Bounds for LXoP[$L, 2, n$]

In this section, we state and prove our main theorem regarding LXoP[$L, 2, n$].

Theorem 3. *Assume that the function $L'(x) = x \oplus L(x)$ is a permutation on \mathbb{F}_2^n . Given that $N \geq 2^{10}$ and $q \leq N/32$,*

$$\text{Opt}_{\text{LXoP}[L,2,n]}^{\text{prf}}(q) \leq \frac{23q}{N^{1.5}},$$

and assuming $\frac{32\sqrt{u}q_{\max}}{N^{1.5}} \leq \frac{1}{2}$,

$$\text{Opt}_{\text{LXoP}[L,2,n],u}^{\text{mu-prf}}(q_{\max}) \leq \frac{32\sqrt{u}q_{\max}}{N^{1.5}}.$$

Throughout this section, we assume that $L'(x) = x \oplus L(x)$ is invertible.

The proof is based on the following two lemmas, proved in the remainder of this section.

Lemma 8. $W^{-1}[\xi_{n,2,q}^{(L)}] = \frac{4q}{(N-1)(N-2)^2}$.

Lemma 9. Let $q \leq N/32$ and $N \geq 2^{10}$. Then $\sum_{k=2}^q W^{-k}[\xi_{n,2,q}^{(L)}] \leq \frac{2^{10.5}q^2}{N^3}$.

Proof (of Theorem 3). By Lemma 8 and Lemma 9,

$$\text{Var}[\xi_{n,2,q}^{(L)}] = \sum_{k=1}^q W^{-k}[\xi_{n,2,q}^{(L)}] \leq \frac{4q}{(N-1)(N-2)^2} + \frac{2^{10.5}q^2}{N^3} \leq \frac{2^{11}q^2}{N^3},$$

as $N \geq 2^{10}$. The result follows from Proposition 7. \blacksquare

6.1 Relation between $\widehat{\xi}_{n,2,k}^{(L)}$ and $\widehat{\mu}_{n,3k}$

We first establish a connection between the Fourier coefficients of $\xi_{n,2,k}^{(L)}$ and those of $\mu_{n,3k}$.

For $\alpha = (\alpha_1, \dots, \alpha_k) = (\alpha_1^1, \alpha_1^2, \dots, \alpha_k^1, \alpha_k^2) \in \widehat{\mathbb{F}}_2^{k \times 2n}$, denote

$$t(\alpha) = (\alpha_1^1, L^T(\alpha_1^2), \alpha_1^2 \oplus L^T(\alpha_1^1), \dots, \alpha_k^1, L^T(\alpha_k^2), \alpha_k^2 \oplus L^T(\alpha_k^1)) \in \widehat{\mathbb{F}}_2^{3k \times n}.$$

Thus $t(\alpha_i) = (\alpha_i^1, L^T(\alpha_i^2), \alpha_i^2 \oplus L^T(\alpha_i^1))$.

Proposition 24 (Relation between $\widehat{\xi}_{n,2,k}^{(L)}$ and $\widehat{\mu}_{n,3k}$). For any $\alpha \in \widehat{\mathbb{F}}_2^{k \times 2n}$, $\widehat{\xi}_{n,2,k}^{(L)}(\alpha) = \widehat{\mu}_{n,3k}(t(\alpha))$.

Proof. By definition of LXoP[$L, 2, n$] and Proposition 2, for any $\alpha \in \widehat{\mathbb{F}}_2^{k \times 2n}$

$$\begin{aligned} \widehat{\xi}_{n,2,k}^{(L)}(\alpha) &= \mathbb{E}_{x \sim \xi_{n,2,k}^{(L)}} [\chi_\alpha(x)] \\ &= \mathbb{E}_{y_{1..k}^{1,2,3} \sim \mu_{n,3k}} [\chi_{\alpha_{1..k}^{1,2}}(y_1^1 \oplus L(y_1^2), y_1^2 \oplus L(y_1^3), \dots, y_k^1 \oplus L(y_k^2), y_k^2 \oplus L(y_k^3))] \\ &= \mathbb{E}_{y_{1..k}^{1,2,3} \sim \mu_{n,3k}} [\chi_{\alpha_{1..k}^1}(y_{1..k}^1) \chi_{L^T(\alpha_{1..k}^2) \oplus \alpha_{1..k}^2}(y_{1..k}^2) \chi_{L^T(\alpha_{1..k}^2)}(y_{1..k}^3)] \\ &= \widehat{\mu}_{n,3k}(\alpha_{1..k}^1, L^T(\alpha_{1..k}^2) \oplus \alpha_{1..k}^2, L^T(\alpha_{1..k}^2)) = \widehat{\mu}_{n,3k}(t(\alpha)). \end{aligned}$$

6.2 Basic Result and Proof of Lemma 8

We prove a basic result and use it to prove Lemma 8.

For $\alpha \in \widehat{\mathbb{F}}_2^{k \times 2n}$, recall that $\#\alpha = |\{i \in [k] \mid \alpha_i \neq 0\}|$ is the size of the support of α (over elements of $\widehat{\mathbb{F}}_2^{2n}$). On the other hand, for $t(\alpha) \in \widehat{\mathbb{F}}_2^{3k \times n}$, $\#t(\alpha) = |\{i \in [3k] \mid t(\alpha)_i \neq 0\}|$. In addition, note that for $i \in [k]$, $\#t(\alpha_i) \in \{0, 1, 2, 3\}$. In fact, as proved below, $\#t(\alpha_i) \in \{0, 2, 3\}$.

Proposition 25 (Basic properties of $t(\alpha)$). For $\alpha = (\alpha^1, \alpha^2) \in \widehat{\mathbb{F}}_2^{2n}$ with $\#\alpha = 1$ (i.e., $\alpha \neq 0$), let $t(\alpha) = (\alpha^1, L^T(\alpha^2), \alpha^2 \oplus L^T(\alpha^1))$.

Then, $\#t(\alpha) \in \{2, 3\}$. Moreover, if $\#t(\alpha) = 2$, then

$$t(\alpha) \in \{(0, L^T(\alpha^2), \alpha^2), (\alpha^1, 0, L^T(\alpha^1)), (\alpha^1, (L^2)^T(\alpha^1), 0)\}.$$

Proof. If $\alpha^1 = 0$, then $t(\alpha) = (0, L^T(\alpha^2), \alpha^2)$, and we have $\alpha^2 \neq 0$ and $L^T(\alpha^2) \neq 0$ (since $t(\alpha) \neq 0$ and $(L)^T$ is a permutation). Hence, $\#t(\alpha) = 2$. Similarly, if $L^T(\alpha^2) = 0$, then $\alpha^2 = 0$ and $\#t(\alpha) = \#(\alpha^1, 0, L^T(\alpha^1)) = 2$. Furthermore if $\alpha^2 \oplus L^T(\alpha^1) = 0$ then $\alpha^2 = L^T(\alpha^1)$ and $\#t(\alpha) = \#(\alpha^1, (L^2)^T(\alpha^1), 0) = 2$ (since $t(\alpha) \neq 0$ and $((L^2)^T)$ is a permutation). ■

We conclude that if $\#\alpha = k$, then $\#t(\alpha) \in [2k, 3k]$. Denote $\#_2\alpha = |\{i \in [k] \mid \#t(\alpha_i) = 2\}|$ and $\#_3\alpha = \#\alpha - \#_2\alpha = |\{i \in [k] \mid \#t(\alpha_i) = 3\}|$. Therefore, if $\#\alpha = k$ and $\#_3\alpha = m$ then $\#t(\alpha) = 2(k - m) + 3m = 2k + m$.

Proof (of Lemma 8). By Proposition 24,

$$W^{-1}[\xi_{n,2,q}^{(L)}] = \binom{q}{1} W^{-1}[\xi_{n,2,1}^{(L)}] = q \sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{2n} \\ \#\alpha=1}} \widehat{\xi}_{n,2,1}^{(L)}(\alpha)^2 = q \sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{2n} \\ \#\alpha=1}} \widehat{\mu}_{n,3}(t(\alpha))^2.$$

Let $\alpha = (\alpha^1, \alpha^2) \in \widehat{\mathbb{F}}_2^{2n}$, hence $t(\alpha) = (\alpha^1, L^T(\alpha^2), \alpha^2 \oplus L^T(\alpha^1))$. By Proposition 9, we have $\widehat{\mu}_{n,3}(t(\alpha))^2 \neq 0$ only if $\alpha^1 \oplus L^T(\alpha^2) \oplus \alpha^2 \oplus L^T(\alpha^1) = 0$. In this case, $\alpha^1 \oplus L^T(\alpha^1) = \alpha^2 \oplus L^T(\alpha^2)$ and thus $(L')^T(\alpha^1) = (L')^T(\alpha^2)$. Hence, by invertibility of $(L')^T$, $\alpha^1 = \alpha^2$, which implies that $t(\alpha) = (\alpha^1, L^T(\alpha^1), \alpha^1 \oplus L^T(\alpha^1))$. In particular, since L^T and $(L')^T$ are invertible then $\#t(\alpha) = 3$.

By Proposition 10, for every $\alpha^1 \neq 0$, $|\widehat{\mu}_{n,3}(\alpha^1, L^T(\alpha^1), \alpha^1 \oplus L^T(\alpha^1))| = \frac{2}{N-2} \frac{1}{N-1}$. Since $\alpha^1 \in \widehat{\mathbb{F}}_2^n$ can attain $N - 1$ non-zero values, we conclude that

$$W^{-1}[\xi_{n,2,q}^{(L)}] = q \sum_{\substack{\alpha \in \widehat{\mathbb{F}}_2^{2n} \\ \#\alpha=1}} \widehat{\mu}_{n,3}(t(\alpha))^2 = q(N - 1) \frac{4}{(N-1)^2(N-2)^2} = \frac{4q}{(N-1)(N-2)^2}.$$

■

6.3 Application of Main Framework and Proof of Lemma 9

We apply our main framework and use it to prove Lemma 9.

Proposition 26. Let $2 \leq k \leq q \leq N/32$, and define $c_k = 0$ if k is even and $c_k = \frac{1}{2}$ otherwise (i.e. $c_k = \frac{k \bmod 2}{2}$). Then

$$W^{=k}[\xi_{n,2,q}^{(L)}] \leq \binom{q}{k} 2^{7k/2+3c_k} \left(\frac{k}{N-2k}\right)^{3k/2+c_k}.$$

Proof. By Proposition 24 and Proposition 25,

$$\begin{aligned} W^{=k}[\xi_{n,2,q}^{(L)}] &= \binom{q}{k} W^{=k}[\xi_{n,2,k}^{(L)}] = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^{2n}} \widehat{\xi}_{n,2,1}^{(L)}(\alpha)^2 = \binom{q}{k} \sum_{\alpha \in \mathcal{M}_{=k,k}^{2n}} \widehat{\mu}_{n,3k}(t(\alpha))^2 \\ &= \binom{q}{k} \sum_{m=0}^k \sum_{\substack{\alpha \in \mathcal{M}_{=k,k}^{2n} \\ \#_3\alpha=m}} \widehat{\mu}_{n,3k}(t(\alpha))^2. \end{aligned}$$

For $m \in \{0, \dots, k\}$, denote $\mathcal{S}^{(k,m)} = \{\alpha \in \widehat{\mathbb{F}}_2^{k \times 2n} \mid \#\alpha = k \wedge \#_3\alpha = m\}$. We have shown that

$$W^{=k}[\xi_{n,2,q}^{(L)}] = \binom{q}{k} \sum_{m=0}^k \sum_{\alpha \in \mathcal{S}^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2. \quad (21)$$

Fix a pair (k, m) . We would like to upper bound $\sum_{\alpha \in \mathcal{S}^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2$ using Lemma 3. For this purpose, according to restriction (a2) of Setting 1, we first need to partition the set $\mathcal{S}^{(k,m)}$ into subsets such that the (transformed) masks in each subset, $t(\alpha)$, share the same non-zero entries (over $\widehat{\mathbb{F}}_2^n$).

We now analyze this partition of $\mathcal{S}^{(k,m)}$. By proposition 25, every $i \in [k]$ with $\#t(\alpha_i) = 2$ has 3 possible structures which determine which 2 of its 3 entries are non-zero over $\widehat{\mathbb{F}}_2^n$. For every $\alpha \in \mathcal{S}^{(k,m)}$, there are $k - m$ such indices i with $\#t(\alpha_i) = 2$. Therefore, there are $\binom{k}{m} 3^{k-m}$ possible non-zero index sets (with non-zero values over $\widehat{\mathbb{F}}_2^n$). Note that every such index set has size equal to $\#t(\alpha) = 2k + m$ for $\alpha \in \mathcal{S}^{(k,m)}$.

Denote by $\Lambda_{k,m}$ the collection of these $\binom{k}{m} 3^{k-m}$ non-zero index sets, where every $\lambda \subseteq [k] \times [3]$ is of size $2k + m$. We thus partition $\mathcal{S}^{(k,m)}$ into $\binom{k}{m} 3^{k-m}$ subsets, denoted $\{\mathcal{S}_\lambda^{(k,m)}\}_{\lambda \in \Lambda_{k,m}}$, each with common non-zero entries of $t(\alpha)$ over $\widehat{\mathbb{F}}_2^n$. Concretely, $\alpha \in \mathcal{S}^{(k,m)}$ satisfies $\alpha \in \mathcal{S}_\lambda^{(k,m)}$ if for every $(i, j) \in [k] \times [3]$, $t(\alpha_i)_j \neq 0$ if and only if $(i, j) \in \lambda$. We have

$$\sum_{\alpha \in \mathcal{S}^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2 = \sum_{\lambda \in \Lambda_{k,m}} \sum_{\alpha \in \mathcal{S}_\lambda^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2. \quad (22)$$

Applying Lemma 3. Fix any $\lambda \in \Lambda_{k,m}$. We now use Lemma 3 to bound $\sum_{\alpha \in \mathcal{S}_\lambda^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2$. For this purpose, let $\mathcal{S} = \mathcal{S}_\lambda^{(k,m)}$ and define $T(\alpha) = T_k(\alpha) = t(\alpha)$ for every $\alpha \in \mathcal{S}_\lambda^{(k,m)}$. In this case, $k' = 3k$ and $k_0 = 2k + m$.

The PIS *pis* resembles the one defined in the proof of Proposition 23. Given a node v , we say that an index $\ell \in [k]$ is redundant if all 3 indices $3\ell - 2, 3\ell - 1$ and 3ℓ are unaltered by Definition 7.

At a given node v , let $\ell \in [k]$ be the largest redundant index. The PIS *pis* will select as primary index the smallest index in the triplet $\{3\ell - 2, 3\ell - 1, 3\ell\}$ that is in \mathcal{NZ}_v .

The recursion is executed up to depth $d = \lceil k/2 \rceil$. As in the proof of Proposition 23, a redundant index is guaranteed to exist up to depth $d - 1 = \lceil k/2 \rceil - 1$ and *pis* is well-defined.

In order to invoke Lemma 3, we prove that the two conditions of Proposition 13 hold. First, by our definition of $\mathcal{S} = \mathcal{S}_\lambda^{(k,m)}$, the pair (\mathcal{S}, T) defined above satisfies the restrictions of Setting 1, and condition (c1) holds.

It remains to prove condition (c2). Specifically, for a node v such that $j = \text{pis}(v)$ and $\alpha \in \mathcal{S}_v$, we prove that $T_v(\alpha)_j$ can be computed from $T_{v_i}(\alpha) = T_v(\alpha)^{\oplus(j,i)}$ (where i is a secondary index).

Note that if ℓ is redundant, then for every $\alpha \in \mathcal{S}_v$, the elements $T_v(\alpha)_{3\ell-2}, T_v(\alpha)_{3\ell-1}, T_v(\alpha)_{3\ell}$ are equal to those of $T(\alpha_\ell) = t(\alpha_\ell)$. Thus, depending on v , according to Proposition 25 we have 4 possibilities for the zero entries of $T_v(\alpha)_{3\ell-2}, T_v(\alpha)_{3\ell-1}, T_v(\alpha)_{3\ell}$.

First, assume that $\#t(\alpha_\ell) = 3$, namely $t(\alpha_\ell) = (\alpha_\ell^1, L^T(\alpha_\ell^2), \alpha_\ell^2 \oplus L^T(\alpha_\ell^1))$ with all 3 entries non-zero. Then the first index with value α_ℓ^1 is selected as primary index ($j = 3\ell - 2$). We need to verify that $T_v(\alpha)_j = \alpha_\ell^1$ can be uniquely recovered from $T_v(\alpha)^{\oplus(j,i)}$ regardless of the secondary index i , namely that it can be recovered from the values of either

$$\begin{aligned} (1) & L^T(\alpha_\ell^2), \alpha_\ell^2 \oplus L^T(\alpha_\ell^1), & \text{in case } i \notin \{3\ell - 1, 3\ell\}, \\ (2) & L^T(\alpha_\ell^2) \oplus \alpha_\ell^1, \alpha_\ell^2 \oplus L^T(\alpha_\ell^1), & \text{in case } i = 3\ell - 1, \text{ or} \\ (3) & L^T(\alpha_\ell^2), \alpha_\ell^2 \oplus L^T(\alpha_\ell^1) \oplus \alpha_\ell^1, & \text{in case } i = 3\ell. \end{aligned}$$

In case (1), α_ℓ^1 can be recovered after computing α_ℓ^2 due to the invertibility of L^T . In case (2), we apply L^T to the second value and XOR to the first to obtain the value of

$$\begin{aligned} (L^2)^T(\alpha_\ell^1) \oplus \alpha_\ell^1 &= (L^2)^T(\alpha_\ell^1) \oplus L^T(\alpha_\ell^1) \oplus L^T(\alpha_\ell^1) \oplus \alpha_\ell^1 \\ &= L^T((L')^T(\alpha_\ell^1)) \oplus (L')^T(\alpha_\ell^1) = ((L')^2)^T(\alpha_\ell^1). \end{aligned}$$

Since $((L')^2)^T$ is invertible, α_ℓ^1 can be uniquely recovered. In case (3), we deduce α_ℓ^2 and then $L^T(\alpha_\ell^1) \oplus \alpha_\ell^1 = (L')^T(\alpha_\ell^1)$, from which we recover α_ℓ^1 since $(L')^T$ is invertible.

Second, if $\#t(\alpha_\ell) = 2$, then according to Proposition 25,

$$t(\alpha_\ell) \in \{(0, L^T(\alpha_\ell^2), \alpha_\ell^2), (\alpha_\ell^1, 0, L^T(\alpha_\ell^1)), (\alpha_\ell^1, (L^2)^T(\alpha_\ell^1), 0)\}.$$

By similar calculation to the case $\#t(\alpha_\ell) = 3$, one can verify that in each of the 3 cases above the first non-zero entry (the value of the primary index) can be recovered from $T_v(\alpha)^{\oplus(j,i)}$ regardless of the secondary index.

We conclude that the two conditions of Proposition 13 hold. Applying our framework of Lemma 3 (with $d = \lceil k/2 \rceil = k/2 + c_k$, $k_0 = 2k + m$), we obtain

$$\sum_{\alpha \in \mathcal{S}_\lambda^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2 \leq 2^d \frac{(k_0)^{2d} (k_0 - 2d)^{k_0/2 - d}}{(N - k_0)^{k_0/2 + d}} = 2^{k/2 + c_k} \frac{(2k+m)^{k+2c_k} (k+m-2c_k)^{k/2+m/2-c_k}}{(N-2k-m)^{3k/2+m/2+c_k}}.$$

We recall that $|A_{k,m}| = \binom{k}{m} 3^{k-m}$ and $\sum_{m=0}^k \binom{k}{m} 3^{k-m} = 4^k$. Using (21), (22) and the inequality above we deduce

$$\begin{aligned} W^{=k}[\xi_{n,2,q}^{(L)}] &= \binom{q}{k} \sum_{m=0}^k \sum_{\alpha \in \mathcal{S}^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2 = \binom{q}{k} \sum_{m=0}^k \sum_{\lambda \in A_{k,m}} \sum_{\alpha \in \mathcal{S}_\lambda^{(k,m)}} \widehat{\mu}_{n,3k}(t(\alpha))^2 \\ &\leq \binom{q}{k} \sum_{m=0}^k |A_{k,m}| 2^{k/2+c_k} \frac{(2k+m)^{k+2c_k} (k+m-2c_k)^{k/2+m/2-c_k}}{(N-2k-m)^{3k/2+m/2+c_k}} \\ &\leq \binom{q}{k} 4^k \max_{m \in \{0,1,\dots,k\}} \left\{ 2^{k/2+c_k} \frac{(2k+m)^{k+2c_k} (k+m)^{k/2+m/2-c_k}}{(N-2k-m)^{3k/2+m/2+c_k}} \right\} \\ &= \binom{q}{k} 2^{5k/2+c_k} \max_{m \in \{0,1,\dots,k\}} \left\{ \frac{(2k+m)^{k+2c_k} (k+m)^{k/2+m/2-c_k}}{(N-2k-m)^{3k/2+m/2+c_k}} \right\}. \end{aligned}$$

Denote $B(m) = \frac{(2k+m)^{k+2c_k}(k+m)^{k/2+m/2-c_k}}{(N-2k-m)^{3k/2+m/2+c_k}}$. Then, assuming $m+1 \leq k \leq q \leq N/32$, by Proposition 1,

$$\begin{aligned} \frac{B(m+1)}{B(m)} &\leq e^{\frac{k+2c_k}{2k+m} + \frac{k/2+m/2-c_k}{k+m} + \frac{3k/2+m/2+c_k}{N-2k-m-1} \left(\frac{k+m+1}{N-2k-m-1}\right)^{1/2}} \\ &\leq e^{1+c_k\left(\frac{2}{2k+m} - \frac{1}{k+m}\right) + \frac{2q}{N-3q} \left(\frac{2q}{N-3q}\right)^{1/2}} \leq e^{1+\frac{c_k}{k} + \frac{2}{29} \left(\frac{2}{29}\right)^{1/2}} \leq e^{\frac{7}{6} + \frac{2}{29} \left(\frac{2}{29}\right)^{1/2}} \leq 1, \end{aligned}$$

where we also used the facts that $k \geq 2$ and $c_2 = 0$, hence $\frac{c_k}{k} \leq \frac{1}{6}$. Therefore,

$$\begin{aligned} W^{=k}[\xi_{n,2,q}^{(L)}] &\leq \binom{q}{k} 2^{5k/2+c_k} \max_{m \in \{0,1,\dots,k\}} B(m) \leq \binom{q}{k} 2^{5k/2+c_k} B(0) \\ &= \binom{q}{k} 2^{5k/2+c_k} \frac{(2k)^{k+2c_k}(k)^{k/2-c_k}}{(N-2k)^{3k/2+c_k}} = \binom{q}{k} 2^{7k/2+3c_k} \left(\frac{k}{N-2k}\right)^{3k/2+c_k}. \end{aligned}$$

■

The proof of Lemma 9 uses Proposition 26 and shows that the bound on $W^{=k}[\xi_{n,2,q}^{(L)}]$ dominates the sum $\sum_{k=2}^q W^{=k}[\xi_{n,2,q}^{(L)}]$.

Proof (of Lemma 9). Applying Proposition 26,

$$\sum_{k=2}^q W^{=k}[\xi_{n,2,q}^{(L)}] \leq \sum_{k=2}^q \binom{q}{k} 2^{7k/2+3+3c_k} \left(\frac{k}{N-2k}\right)^{3k/2+c_k}.$$

Denote $B(k) = \binom{q}{k} 2^{7k/2+3+3c_k} \left(\frac{k}{N-2k}\right)^{3k/2+c_k}$ and note that $c_{k+2} = c_k$. Then, assuming $k+2 \leq q \leq N/32$, and recalling that $k \geq 2$ (and $c_2 = 0$), by Proposition 1,

$$\begin{aligned} \frac{B(k+2)}{B(k)} &\leq \frac{q^2}{(k+1)(k+2)} 2^7 e^{\frac{3k+2c_k}{k} + \frac{2(3k+2c_k)}{N-2k-4}} \left(\frac{k+2}{N-2k-4}\right)^3 \leq 2^7 e^{3+\frac{1}{3} + \frac{6q}{N-2q} \frac{k+2}{k+1} \frac{q^2(k+2)}{(N-2q)^3}} \\ &\leq 2^{7\frac{4}{3}} e^{3+\frac{1}{3} + \frac{6}{30} \frac{q^3}{(30q)^3}} \leq 2^{7\frac{4}{3}} e^4 (30)^{-3} \leq \frac{1}{2}. \end{aligned}$$

Therefore, using the facts that $N \geq 2^{10}$ and $q \leq N/32$,

$$\begin{aligned} \sum_{k=2}^q W^{=k}[\xi_{n,2,q}^{(L)}] &\leq \sum_{k=2}^q B(k) \leq 2B(2) + 2B(3) \leq 2\binom{q}{2} 2^7 \left(\frac{2}{N-4}\right)^3 + 2\binom{q}{3} 2^{12} \left(\frac{3}{N-6}\right)^5 \\ &\leq \frac{2^{10} q^2}{(N-4)^3} + \frac{2^{12} 3^4 q^3}{(N-6)^5} \leq \frac{2^{10.3} q^2}{N^3} + \frac{2^{20} q^2}{N^3} \frac{q}{N} \frac{1}{N} \leq \frac{2^{10.3} q^2}{N^3} + \frac{2^{20} q^2}{N^3} \frac{1}{32} \frac{1}{2^{10}} \leq \frac{2^{10.5} q^2}{N^3}. \end{aligned}$$

■

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A Bounds on advantage for symmetric functions

Proposition 27 (Bounds on advantage for symmetric functions). *Assume that the output distribution generated by $H : \mathcal{K} \times \{0, 1\}^m \mapsto \{0, 1\}^n$ is independent of the queries of the adversary. Denote by $\varphi_{H(n,q)} : \mathbb{F}_2^{q \times n} \rightarrow \mathbb{R}^{\geq 0}$ the density function generated by H . Moreover, assume that $\varphi_{H(n,q)}$ is symmetric in the sense that every element of the sample is marginally distributed as $\varphi_{H(n,1)}$. Then,*

$$\text{Opt}_H^{\text{prf}}(q) \leq q \text{SD}(\varphi_{H(n,1)}, \mathbf{1}_n) + \frac{1}{2} \sqrt{\sum_{k=2}^q W^{=k}[\varphi_{H(n,q)}]}.$$

Proof. Let $\mathcal{S} = \mathcal{M}_{=1,q}^n = \{\alpha \in \widehat{\mathbb{F}}_2^{q \times n} \mid \#\alpha = 1\}$. By (1) and the upper bound of Proposition 5,

$$\begin{aligned} 2\text{Opt}_H^{\text{prf}}(q) &= 2\text{SD}(\varphi_{H(n,q)}, \mathbf{1}_{qn}) \\ &\leq \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\varphi}_{H(n,q)}(\alpha) \chi_\alpha(x) \right| + \sqrt{\sum_{k=2}^q \mathbb{W}^{=k}[\varphi_{H(n,q)}]}. \end{aligned}$$

It remain to prove that

$$\mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\varphi}_{H(n,q)}(\alpha) \chi_\alpha(x) \right| \leq 2q \text{SD}(\varphi_{H(n,1)}, \mathbf{1}_n).$$

For $\alpha \in \widehat{\mathbb{F}}_2^{q \times n}$ with $\#\alpha = 1$, define $\text{in}(\alpha)$ to be the unique index i with $\alpha_i \neq 0$. By symmetry of $\varphi_{H(n,q)}$, we have $\widehat{\varphi}_{H(n,q)}(\alpha) = \widehat{\varphi}_{H(n,1)}(\alpha_{\text{in}(\alpha)})$. Therefore,

$$\begin{aligned} &\mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\varphi}_{H(n,q)}(\alpha) \chi_\alpha(x) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\varphi}_{H(n,1)}(\alpha_{\text{in}(\alpha)}) \prod_{i \in [q]} \chi_{\alpha_i}(x_i) \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{\alpha \in \mathcal{M}_{=1,q}^n} \widehat{\varphi}_{H(n,1)}(\alpha_{\text{in}(\alpha)}) \chi_{\alpha_{\text{in}(\alpha)}}(x_{\text{in}(\alpha)}) \right| \\ &= \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{i=1}^q \sum_{\substack{\beta \in \widehat{\mathbb{F}}_2^n \\ \beta \neq 0}} \widehat{\varphi}_{H(n,1)}(\beta) \chi_\beta(x_i) \right| = \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} \left| \sum_{i=1}^q (\varphi_{H(n,1)}(x_i) - \widehat{\varphi}_{H(n,1)}(0) \chi_0(x_i)) \right| \\ &\leq \sum_{i=1}^q \mathbb{E}_{x \sim \widehat{\mathbb{F}}_2^{q \times n}} |(\varphi_{H(n,1)}(x_i) - 1)| = q \mathbb{E}_{y \sim \widehat{\mathbb{F}}_2^n} |\varphi_{H(n,1)}(y) - 1| = 2q \text{SD}(\varphi_{H(n,1)}, \mathbf{1}_n). \end{aligned}$$

■