

The Name of the Title Is Hope

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ABSTRACT

tbd.

CCS CONCEPTS

• Theory of computation → Cryptographic primitives.

KEYWORDS

tbd

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1 INTRODUCTION

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2 PRELIMINARY

2.1 Basic Notations

Point and multi-point functions. Given a domain size N and Abelian group \mathbb{G} , a point function $f_{\alpha,\beta} : [N] \rightarrow \mathbb{G}$ for $\alpha \in [N]$ and $\beta \in \mathbb{G}$ evaluates to β on input α and to $0 \in \mathbb{G}$ on all other inputs. We denote by $\hat{f}_{\alpha,\beta} = (N, \mathbb{G}, \alpha, \beta)$ the representation of such a point function. A multi-point function $f_{A,B} : [N] \rightarrow \mathbb{G}$ for $A = (\alpha_1, \dots, \alpha_t) \in [N]^t$ and $B = (\beta_1, \dots, \beta_t) \in \mathbb{G}^t$ evaluates to β_i on input α_i for $1 \leq i \leq t$ and to 0 on all other inputs. Denote $\hat{f}_{A,B}(N, \mathbb{G}, A, B)$ the representation of such a point function.

Enote: MPF. Also representation of groups.

2.2 Distributed Multi-Point Functions

Enote: should directly adapt to multi-point function case

We begin by defining a slightly generalized notion of distributed point functions (DPFs), which accounts for the extra parameter \mathbb{G}' .

DEFINITION 1 (DPF [1, 3]). A (2-party) distributed point function (DPF) is a triple of algorithms $\Pi = (\text{Gen}, \text{Eval}_0, \text{Eval}_1)$ with the following syntax:

- $\text{Gen}(1^\lambda, \hat{f}_{\alpha,\beta}) \rightarrow (k_0, k_1)$: On input security parameter $\lambda \in \mathbb{N}$ and point function description $\hat{f}_{\alpha,\beta} = (N, \mathbb{G}, \alpha, \beta)$, the (randomized) key generation algorithm Gen returns a pair of keys $k_0, k_1 \in \{0, 1\}^*$. We assume that N and \mathbb{G} are determined by each key.

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- $\text{Eval}_i(k_i, x) \rightarrow y_i$: On input key $k_i \in \{0, 1\}^*$ and input $x \in [N]$ the (deterministic) evaluation algorithm of server i , Eval_i returns $y_i \in \mathbb{G}$.

We require Π to satisfy the following requirements:

- **Correctness:** For every λ , $\hat{f} = \hat{f}_{\alpha,\beta} = (N, \mathbb{G}, \alpha, \beta)$ such that $\beta \in \mathbb{G}$, and $x \in [N]$, if $(k_0, k_1) \leftarrow \text{Gen}(1^\lambda, \hat{f})$, then

$$\Pr \left[\sum_{i=0}^1 \text{Eval}_i(k_i, x) = f_{\alpha,\beta}(x) \right] = 1$$

- **Security:** Consider the following semantic security challenge experiment for corrupted server $i \in \{0, 1\}$:

- (1) The adversary produces two point function descriptions $(\hat{f}^0 = (N, \mathbb{G}, \alpha_0, \beta_0), \hat{f}^1 = (N, \mathbb{G}, \alpha_1, \beta_1)) \leftarrow \mathcal{A}(1^\lambda)$, where $\alpha_i \in [N]$ and $\beta_i \in \mathbb{G}$.
- (2) The challenger samples $b \leftarrow \{0, 1\}$ and $(k_0, k_1) \leftarrow \text{Gen}(1^\lambda, \hat{f}^b)$.
- (3) The adversary outputs a guess $b' \leftarrow \mathcal{A}(k_i)$.

Denote by $\text{Adv}(1^\lambda, \mathcal{A}, i) = \Pr[b = b'] - 1/2$ the advantage of \mathcal{A} in guessing b in the above experiment. For every non-uniform polynomial time adversary \mathcal{A} there exists a negligible function ν such that $\text{Adv}(1^\lambda, \mathcal{A}, i) \leq \nu(\lambda)$ for all $\lambda \in \mathbb{N}$.

We will also be interested in applying the evaluation algorithm on all inputs. Given a DPF $(\text{Gen}, \text{Eval}_0, \text{Eval}_1)$, we denote by FullEval_i an algorithm which computes Eval_i on every input x . Hence, FullEval_i receives only a key k_i as input.

2.3 Batch Codes

combinatorial/probabilistic batch codes, with cuckoo hashing a concrete instantiation

2.4 Oblivious Key-Value Stores

DEFINITION 2 (OKVS[2, 4]). An Oblivious Key-Value Stores (OKVS) scheme is a pair of randomized algorithms $(\text{Encoder}, \text{Decoder})$ with respect to a statistical security parameter λ_{stat} and a computational security parameter λ , a randomness space $\{0, 1\}^k$, a key space \mathcal{K} , a value space \mathcal{V} , input length n and output length m . The algorithms are of the following syntax:

- $\text{Encoder}_r(\{(k_1, v_1), (k_2, v_2), \dots, (k_n, v_n)\}) \rightarrow P$: On input n key-value pairs with distinct keys, the encode algorithm with randomness r in the randomness space outputs an encoding $P \in \mathcal{V}^m \cup \perp$.
- $\text{Decoder}_r(P, k) \rightarrow v$: On input a (nonempty) encoding from \mathcal{V}^m and a key $k \in \mathcal{K}$, output a value v .

We require the scheme to satisfy

- **Correctness:** For every $S \in (\mathcal{K} \times \mathcal{V})^n$, $\Pr_{r \leftarrow \{0, 1\}^k} [\text{Encoder}_r(S) = \perp] \leq 2^{-\lambda_{\text{stat}}}$.

- **Obliviousness:** For any distinct key sets $\{k_1^0, k_2^0, \dots, k_n^0\}$ and $\{k_1^1, k_2^1, \dots, k_n^1\}$ that are different, if they are paired with random values then their encodings are computationally indistinguishable, i.e.,

$$\{r, \text{Encode}_r(\{(k_1^0, v_1), \dots, (k_n^0, v_n)\})\}_{v_1, \dots, v_n \leftarrow \mathcal{V}, r \leftarrow \{0,1\}^\kappa} \\ \approx_c \{r, \text{Encode}_r(\{(k_1^1, v_1), \dots, (k_n^1, v_n)\})\}_{v_1, \dots, v_n \leftarrow \mathcal{V}, r \leftarrow \{0,1\}^\kappa}$$

One can obtain a linear OKVS if in addition require:

- **Linearity:** There exists a function family $\{\text{row}_r : \mathcal{K} \rightarrow \mathcal{V}^m\}_{r \in \{0,1\}^\kappa}$ such that $\text{Decode}_r(P, k) = \langle \text{row}_r(k), P \rangle$.

The Encode process for a linear OKVS is the process of sampling a random P from the set of solutions of the linear system $\{\langle \text{row}_r(k_i), P \rangle = v_i\}_{1 \leq i \leq n}$.

We evaluate an OKVS scheme by its encoding size (output length m), encoding time and decoding time. We stress the following two (linear) OKVS constructions:

CONSTRUCTION 1 (POLYNOMIAL). Suppose $\mathcal{K} = \mathcal{V} = \mathbb{F}$ is a field. Set

- $\text{Encode}(\{(k_i, v_i)\}_{1 \leq i \leq n}) \rightarrow P$ where P is the coefficients of a $(n-1)$ -degree \mathbb{F} -polynomial g_P that $g_P(k_i) = v_i$ for $1 \leq i \leq n$.
- $\text{Decode}(P, k) \rightarrow g_P(k)$.

The polynomial OKVS possesses optimal encoding size, but the Encode process is a polynomial interpolation which is only known to be achieved in super linear time.

CONSTRUCTION 2 (3H-GCT[2, 4]). Suppose $\mathcal{V} = \mathbb{F}$ is a field. Set $\text{row}_r(k) := \text{row}_r^{\text{sparse}}(k) \parallel \text{row}_r^{\text{dense}}(k)$ where $\text{row}_r^{\text{sparse}}$ outputs a uniformly random vector in $\{0,1\}^{m_1}$ of hamming weight 3, and $\text{row}_r^{\text{dense}}(k)$ outputs a short dense vector in \mathbb{F}^{m_2} .

- $\text{Encode}(\{(k_i, v_i)\}_{1 \leq i \leq n}) \rightarrow P$ where P is solved from the system $\{\langle \text{row}_r(k_i), P \rangle = v_i\}_{1 \leq i \leq n}$ using the triangulation algorithm in [4].
- $\text{Decode}(P, k) \rightarrow \langle \text{row}_r(k), P \rangle$.

TBD: mention some connections to cuckoo hashing

3 NEW DMPF CONSTRUCTIONS

3.1 Big-State DMPF

display the big-state DMPF (plus distributed gen)

3.2 Batch-Code DMPF

display the batch-code DMPF

3.3 OKVS-based DMPF

display the OKVS-based DMPF (plus distributed gen)

3.4 Comparison

Comparison table dependent to PRG & F-MUL(list the formulas?)
analyze tradeoff
distributed gen advantage

4 APPLICATIONS

4.1 PCG for OLE from Ring-LPN

Characterize parameters

show nonregular optimization

plug in new DMPF and show overall optimization

4.2 PSI-WCA

plug in new DMPF and analyze advantage interval

plug in distributed gen

4.3 Heavy-hitters

private heavy-hitter

or parallel ORAM?

5 ACKNOWLEDGMENTS

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A BATCH-CODE DMPF SCHEME

B SECURITY PROOFS