# Oblivious Polynomial Evaluation and Secure Set-Intersection from Algebraic PRFs TCC 2015

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## Pseudorandom Functions (PRFs)

PRF is indistinguishable from a random function

#### PRF:

Let  $k \leftarrow \{0,1\}^n$ Given x output  $F_k(x)$   $\simeq$ 

#### **Random function:**

Let R: {0,1}<sup>n</sup> → {0,1}<sup>n</sup> Given x output R(x)

Adversary makes polynomial number of queries

PRFs with Abelian group range

 Support fast batch PRFs evaluation given the PRF key

 Originally introduced in the context of verifiable delegation of polynomial evaluation

#### **Closed from efficiency:**

$$CFEval_{h,z}(x,k)=\prod_{i=0}^{l}[PRF_{k}(z_{i})]^{h_{i}(x)}$$
  
Running time of CFeval is **sublinear** in I

Special case: z = (0,...,d) and  $h_i(x)=x^i$ 

- Implementations in prime order groups:
  - Strong DDH:

$$\begin{aligned} \left| \Pr \left[ D\left( G, p, g, g^x, g^{x^2}, \dots, g^{x^d} \right) \right] \\ &- \left[ D(G, p, g, g^{x_1}, g^{x_2}, \dots, g^{x_d}) \right] \right| \leq negl(\cdot) \\ &\text{PRF}_{k}(x) = g^{k_0 k_1^x} \end{aligned}$$

• **DDH** (Naor-Reingold):

$$PRF_k(x_1,...,x_m) = g^{k_0 \prod_{i=1}^m k_i^{x_i}}$$

#### Strong DDH

CEEval with constant overhead follows from the identity:

$$\sum_{i=0}^{d} k_0 k_1^i x^i = \frac{k_0 (k_1^{d+1} x^{d+1} - 1)}{k_1 x - 1}$$

#### DDH [NR]

CEEval with log d overhead follows from the identity:

$$\sum_{j=0}^{d} k_0 \prod_{i=1}^{j+1} k_i^{x_i} = k_0 (1+k_1 x)(1+k_2 x^2) \cdots (1+k_{\log d} x^d)$$

# Verifiable Polynomial Evaluation [BGV11]

**Setup:** The client stores  $\mathbf{Q}(\cdot) = (\mathbf{q}_0, ..., \mathbf{q}_d)$  together with  $(\mathbf{F}_k(\mathbf{0})\mathbf{g}^{\mathsf{aq_0}}, ..., \mathbf{F}_k(\mathbf{d})\mathbf{g}^{\mathsf{aq_d}}) = (\mathbf{g}^{\mathsf{r_0 + aq_0}}, ..., \mathbf{g}^{\mathsf{r_d + aq_d}})$ 

**Evaluation:** Given **x**, the server returns  $\mathbf{Q}(\mathbf{x})$  and  $\mathbf{g}^{\sum_i (\mathbf{r_i} + a\mathbf{q_i})^{\mathbf{x^i}}}$ 

**<u>Verification:</u>** Given  $\mathbf{Q}(\mathbf{x})$  and proof  $\mathbf{w}$  check that

$$\mathbf{w} = \mathbf{g}^{\sum_{i} \mathbf{r}^{\mathbf{x}^{i}} + a\mathbf{Q}(\mathbf{x})}$$

- $\circ$  Computing  $\textbf{g}^{\sum_i r_i^{x^i}}$  requires sublinear time in d given the PRF key
- The protocol is not private

### **Oblivious Polynomial Evaluation (OPE)**

Sender

Receiver





Inputs:	$Q(\cdot)=(q_0,q_d)$	X
Outputs:		Q(x)

### **Oblivious Polynomial Evaluation (OPE)**

- Very useful building block:
  - RSA key generation [Gil99]
  - Approximation of Taylor series [LP02]
  - Set-intersection [FNP04]
  - Oblivious keyword search [FIPR05]
  - Secure equality of strings [NP06]
  - Data entanglement [ADDV12]

Prior maliciously secure work uses cut-and-choose or somewhat homomorphic SIMD approach

## **OPE from Algebraic PRFs**

#### Our results:

OPE in the exponent with malicious security

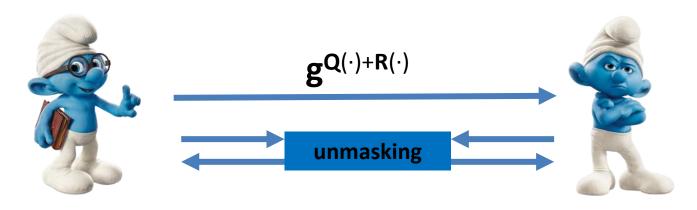
#### Two phases protocol:

- 1. Use algebraic PRFs to mask the sender's polynomial
- 2. Unmask the evaluated masked polynomial

### **OPE from Algebraic PRFs**

Masking: Given input  $g^{Q(\cdot)}=(g^{q_0},...g^{q_d})$  the sender sends the masked polynomial  $g^{Q(\cdot)+R(\cdot)}=(g^{q_0+r_0},...,g^{q_d+r_d})$ 

<u>Unmasking:</u> The receiver computes  $g^{Q(x)+R(x)}$  and the parties compute  $g^{R(x)}$ 



Sender Receiver

## **OPE from Algebraic PRFs**

o Unmasking requires computing  $\mathbf{g}^{\sum_i \mathbf{r_i}^{\mathbf{x^i}}}$  which can be carried out in **sublinear time in d** 

#### Overhead:

- d+1 exponentiations in masking phase
- O(1)/O(log d) exponentiations in unmasking phase

#### **Secure Set-Intersection**

Sender



#### Receiver



Inputs:	X	Y
<b>Outputs:</b>		X∩Y

#### **Secure Set-Intersection**

 Intensively studied due to many applications [FNP04,KS05,DSMRY09,JL09,JL10,HL10,HN12]

- Two common approaches in the plain model:
  - Oblivious polynomial evaluation [FNP04]
  - Oblivious PRF evaluation [FIPR05,LP10]

### **Set-Intersection - The OPE Approach**

- 1. On input  $(x_1,...,x_m)$ , sender masks polynomial  $Q(t)=(x_1-t)\cdots(x_m-t)$  and sends  $g^{Q(t)+R(t)}$
- 2. For each **y**∈**Y** 
  - Receiver evaluates g<sup>Q(y)+R(y)</sup>
  - The parties "unmask" the result by verifying whether  $g^{Q(y)+R(y)} = g^{R(y)}$
- Overhead is |X|·|Y|
  - Reduce overhead using hash functions

# Set-Intersection - The OPE Approach and Hash Functions

- Split the set into bins
  - Receiver evaluates a small degree poly each time
  - Easy to evaluate a polynomial of a particular bin

#### Ocorrectness:

- Sender proves that the polynomials are correct, potentially can use zero polynomials or too many elements
- Receiver proves that it uses the same input when more than one hash is used

# Set-Intersection - The OPE Approach and Hash Functions

#### Overhead:

O(m<sub>x</sub>+m<sub>y</sub>loglog m<sub>x</sub>) under d-strong assumption

•  $O(m_X + m_Y log m_X)$  under **DDH** assumption Better than [HN12]!

# Set-Intersection - The Oblivious PRF Approach

The oblivious PRF functionality:

Sender Receiver





Inputs:	k	X
Outputs:		F <sub>k</sub> (x)

# Set-Intersection - The Oblivious PRF Approach

- First phase: sender picks PRF key k and sends  $F_{k}(x)$  for all  $x \in X$
- Second phase: parties run oblivious PRF, receiver learns F<sub>κ</sub>(y) for all y ∈ Y

Requires committed oblivious PRF!

# Set-Intersection - The Committed Oblivious PRF Approach

The committed oblivious PRF functionality:

Sender Receiver





Inputs:	k	<b>X</b> <sub>1</sub> ,, <b>X</b> <sub>n</sub>
Outputs:		$F_k(x_1),, F_k(x_n)$

# Set-Intersection - The Committed Oblivious PRF Approach

- Algebraic PRFs easily imply committed PRFs
  - Use same protocols for unmasking

- One particular example is a simple committed Naor-Reingold PRF
  - Prior to that, no simple committed protocol

# Set-Intersection - The Committed Oblivious PRF Approach

#### o Overhead:

O(m<sub>X</sub>+m<sub>Y</sub>) under strong-DDH assumption
 Same as [JL09] but for prime order groups
 No setup (CRS)
 Proof complexity is independent of PRF domain

•  $O((m_X + m_Y))log(m_X + m_Y))$  under DDH assumption

#### **Future Research**

- Use algebraic PRFs for more applications
- Construct new algebraic PRFs for large domains

- Recent related work:
  - Aggregate Pseudorandom Functions and Connections to Learning
    - Aloni Cohen and Shafi Goldwasser and Vinod Vaikuntanathan