

Trapdoor Hash Functions and Their Applications

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Setting: Sender-Receiver Computation



Sender

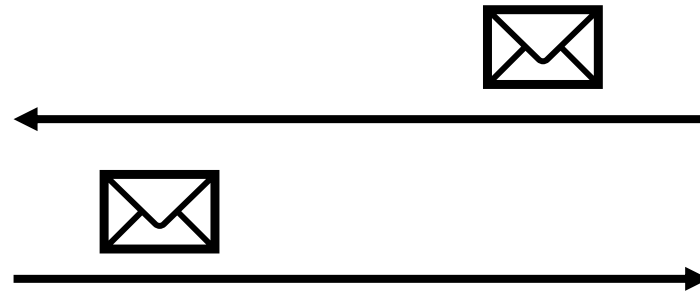
$y \in Y$

$$f: X \times Y \rightarrow Z$$



Receiver

$x \in X$



communication

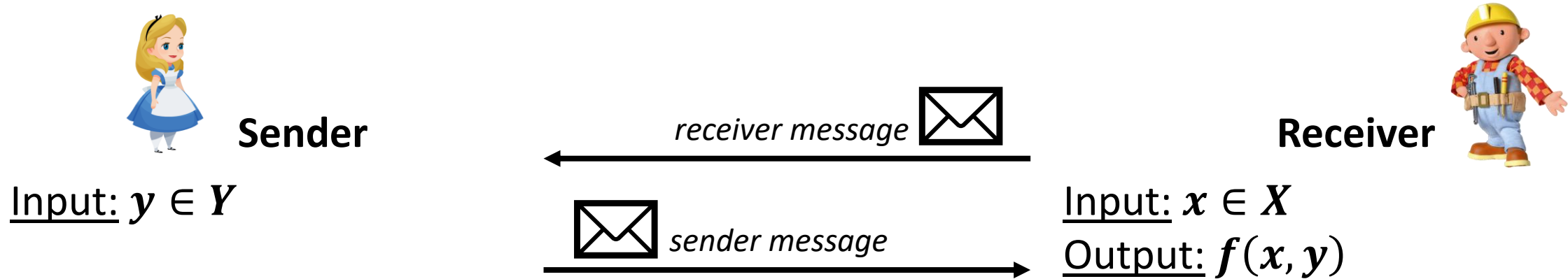


Output:

$f(x, y)$

Setting: Sender-Receiver Computation

Focus: **two-message** protocols with **minimum communication**.



Ideal World:

Simple and **optimal** solutions.



Real World (**semi-honest security**):

Sender and receiver do not trust each other, and want to keep their inputs **private**.

Main question in this work:

Can Secure Protocols be as Efficient as Ideal-World Solutions?

Can Secure Protocols be as Efficient as Ideal-World Solutions?

Setting I.

Sender input is larger than receiver input

$$|y| \gg |x|$$

Example:

String Oblivious Transfer (OT)



$y_0, y_1 \in \{0,1\}^n$



$x \in \{0,1\}$



Output: y_x

In Ideal World, communication dominated by length of **second message** $|f(x, y)| = n$.

Goal

Optimize length of **second message**, i.e.

$$\text{download rate} = \frac{|f(x, y)|}{|\text{Second Message}|}$$

Ideal-World Solution:

download rate = 1

Negative Answer

Secure protocols with **exact** rate 1 **do not exist.**
(even with correlated randomness)

For 2^λ security,

$$|\text{second message}| > n + 2\lambda$$

Best we can hope for:

download rate $\rightarrow 1$

when $n \rightarrow \infty$.

Positive Answer

Protocols with rate approaching 1 using
Trapdoor Hash.

Results in **Setting I**: Rate-1 Oblivious Transfer and More

Life before Trapdoor Hash

- **Rate - ½ OT** is easy to achieve.
- Only known solution for higher rate uses **high-rate homomorphic encryption**.
- Only rate-1 homomorphic encryption scheme: **Damgård-Jurik** scheme based on **DCR**.

Exception: [Gentry-Halevi'19]

Oblivious Transfer (OT)

DCR	LWE	QR	DDH
rate-1 [DJ00]	rate-½	rate-½	rate-½

Life after Trapdoor Hash

Oblivious Transfer (OT)

DCR	LWE	QR	DDH
rate-1	rate-1	rate-1	rate-1

Oblivious Linear-Function Evaluation (OLE)

rate-1	rate-1	rate-1	rate-1
--------	--------	--------	--------

Oblivious Matrix-Vector Product (OMV)

-	rate-1	rate-1	-
---	--------	--------	---

Results in **Setting I**: Application of Rate-1 OT

1. Single-Server Private Information Retrieval [Kushilevitz-Ostrovsky'97]

First **rate-1** PIR protocols with **polylogarithmic communication** [Ishai-Paskin'07].

	DCR	LWE	QR	DDH
Before TDH	$O(\log^c n)$, rate-1	$O(\log^c n)$	$O(2^{\sqrt{\log n}})$ [KO97]	$O(2^{\sqrt{\log n}})$ [KO97]
After TDH	$O(\log^c n)$, rate-1	$O(\log^c n)$, rate-1	$O(\log^c n)$, rate-1	$O(\log^c n)$, rate-1

2. Homomorphic Encryption for Branching Programs

First **semi-compact** homomorphic encryption from **DDH,QR** [Ishai-Paskin'07].

3. Lossy Trapdoor Functions [Peikert-Waters'07]

First **rate-1** constructions.

	DCR	LWE	QR	DDH
Before TDH	rate-1	-	-	constant rate [GGH18]
After TDH	rate-1	rate-1	rate-1	rate-1

Can Secure Protocols be as Efficient as Ideal-World Solutions?

Setting I.

Sender input is larger than receiver input

$$|y| \gg |x|$$

Example:

String Oblivious Transfer (OT)



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Setting II.

Receiver input is larger than sender input

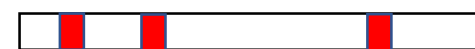
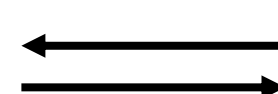
$$|y| \ll |x|$$

Example:

RAM Computation on Big Data



RAM machine M
w/ running time
 $T \ll n$



$$x \in \{0,1\}^n$$

Output: $M(x)$



(DNA analysis, suspects lookup, etc.)

Can Secure Protocols be as Efficient as Ideal-World Solutions?

Setting I.

Sender input is larger than receiver input

$$|y| \gg |x|$$

Example:

String Oblivious Transfer (OT)



$$y_0, y_1 \in \{0,1\}^n$$



$$x \in \{0,1\}$$

Output: y_x



In Ideal World, communication dominated by length of **second message** $|f(x, y)| = n$.

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$$\text{download rate} = \frac{|f(x, y)|}{|\text{Second Message}|}$$

Setting II.

Receiver input is larger than sender input

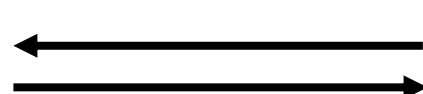
$$|y| \ll |x|$$

Example:

RAM Computation on Big Data



RAM machine M
w/ running time
 $T \ll n$



$x \in \{0,1\}^n$
Output: $M(x)$



In Ideal World, communication is $|y| \ll n$:
"much smaller" than n

Goal

Communication **smaller than $|x| = n$** .
Non-trivial for **two-message** protocols.

Results in **Setting II**: Sublinear Secure RAM Computation

Life before Trapdoor Hash

Only **fully secure** solution is based on **lattice assumptions**.

	Overall Com.	Assumption	Security
Laconic Function Evaluation [QWW18]	$\tilde{O}(T)$	LWE	full
Laconic OT [CDGGMP17]	$\tilde{O}(T)$	DDH	UMA

Life after Trapdoor Hash

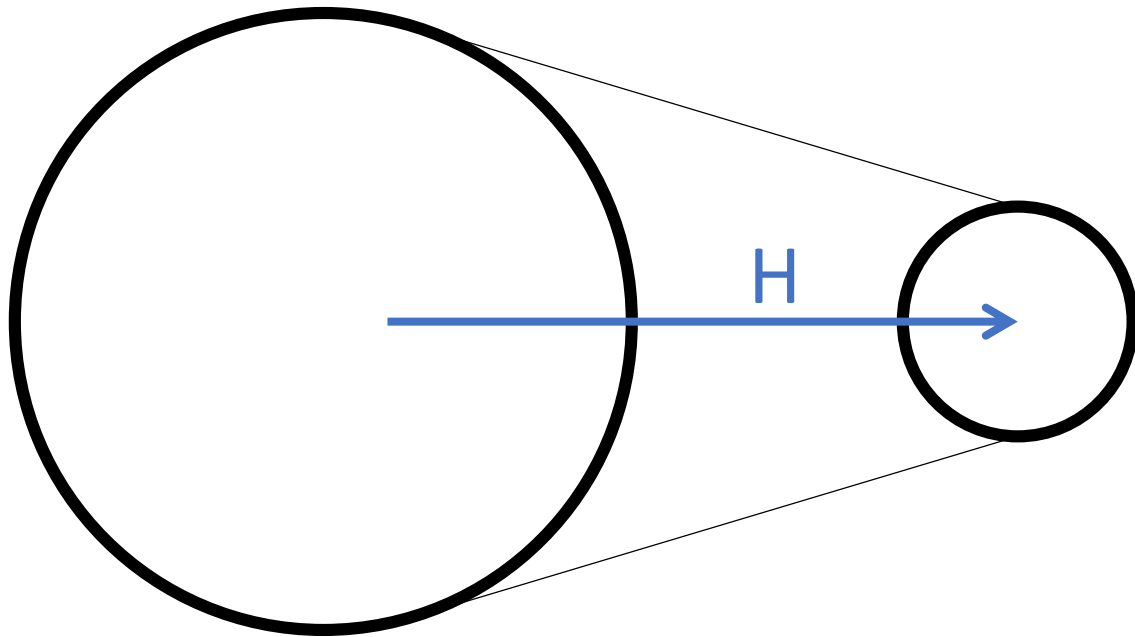
First sublinear two-message protocols under **number-theoretic assumptions**.

Private Laconic OT (through TDH)	$O(T \cdot \sqrt{n})$	DDH	full
	$O(T \cdot \sqrt[3]{n})$	SXDH+	full

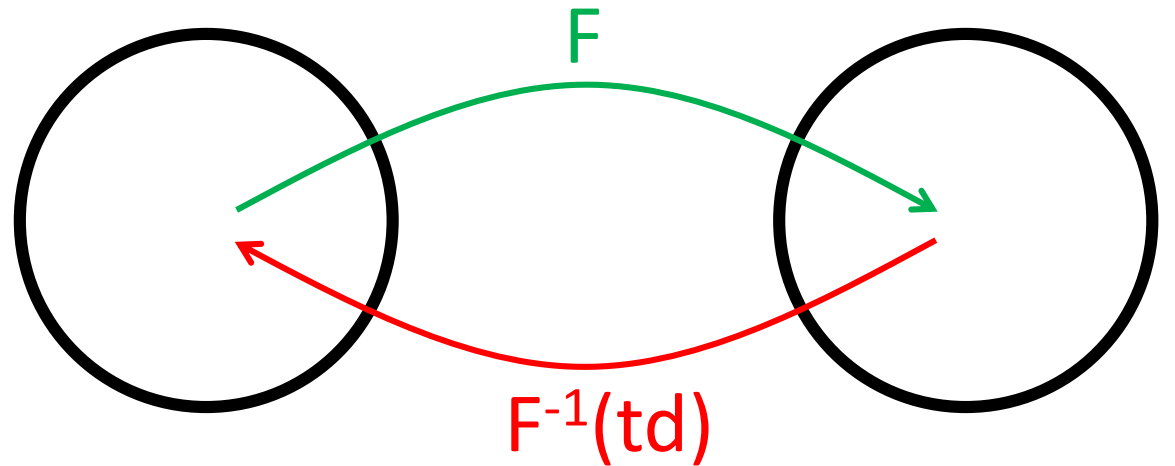
Open Question (already): can we close the efficiency gap?

Trapdoor Hash Functions

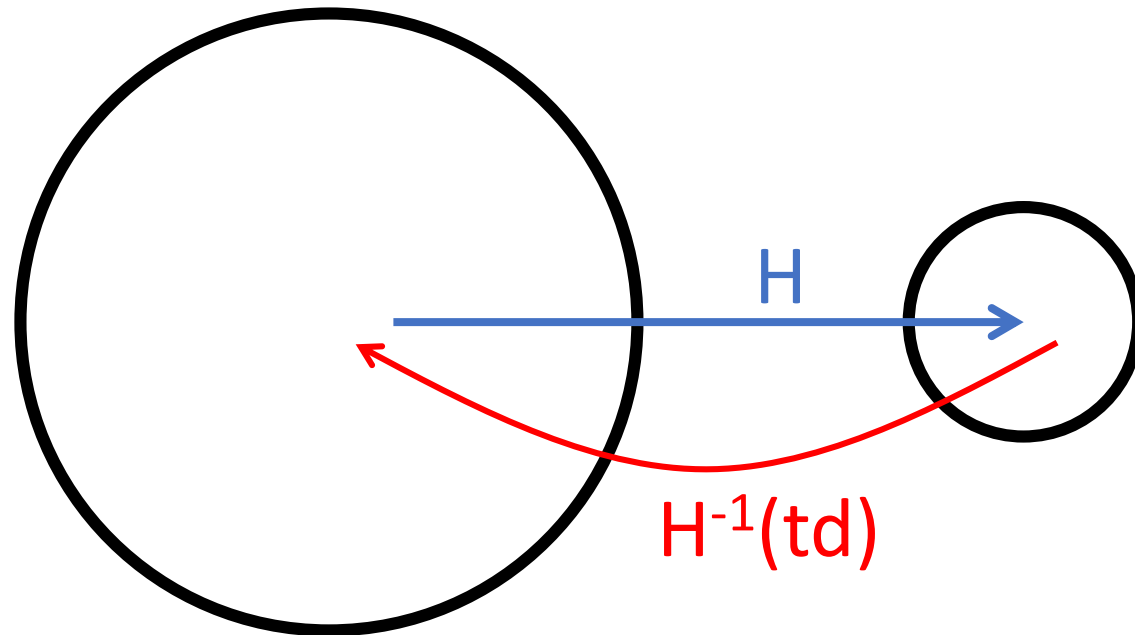
Hash Functions



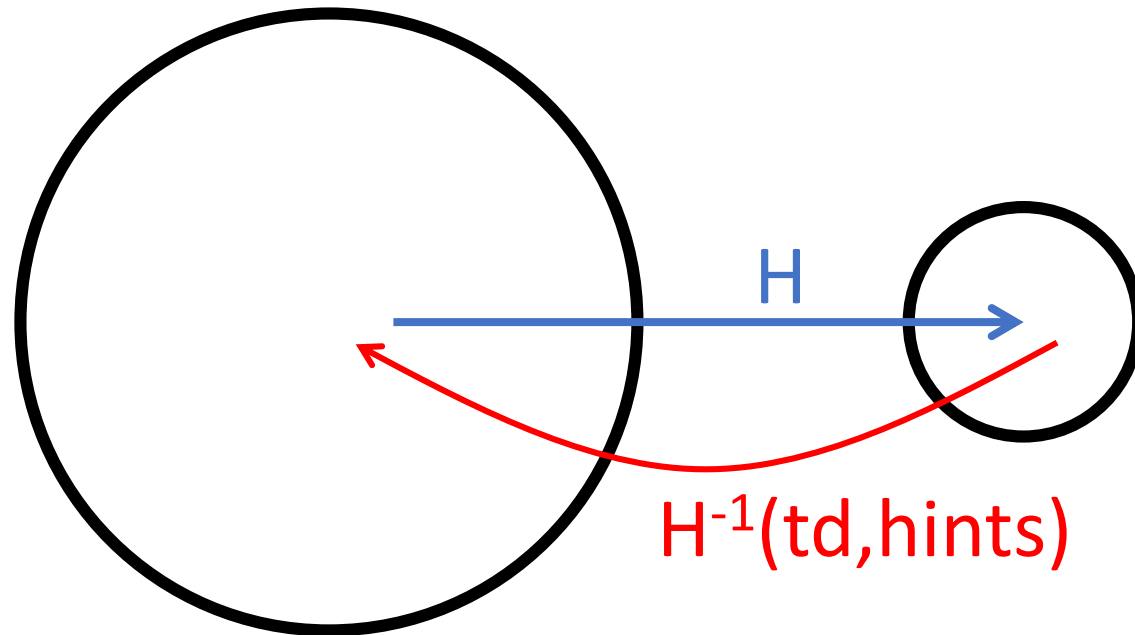
Trapdoor Functions



Trapdoor Hash Functions



Trapdoor Hash Functions



Defining Trapdoor Hash

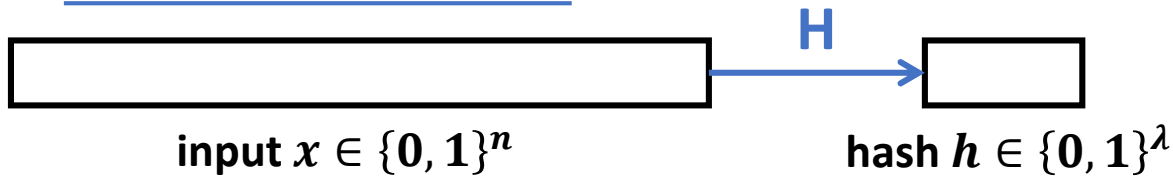


Alice



Bob

The Hash Function:

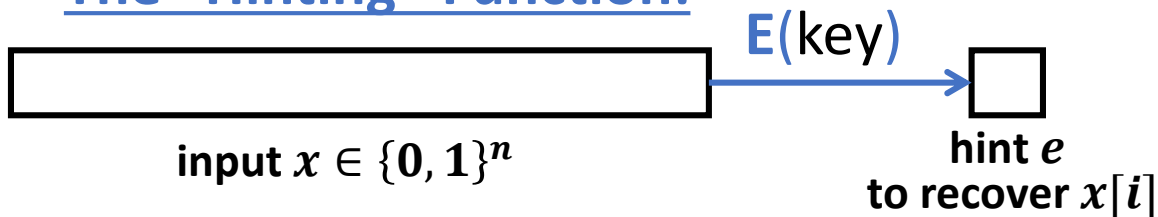


Input Privacy: h hides x .

h

“I want to learn $x[i]$, give me some hints”

The “Hinting” Function:



Efficiency: small hints, i.e. **high rate**

Key Generation:

$(\text{key}, \text{trapdoor}) \leftarrow \mathbf{G}(i)$

Index Privacy: key hides i .

key

Decoding:

$x[i] \leftarrow \mathbf{D}(\text{trapdoor}, h, e)$

e

Defining Trapdoor Hash

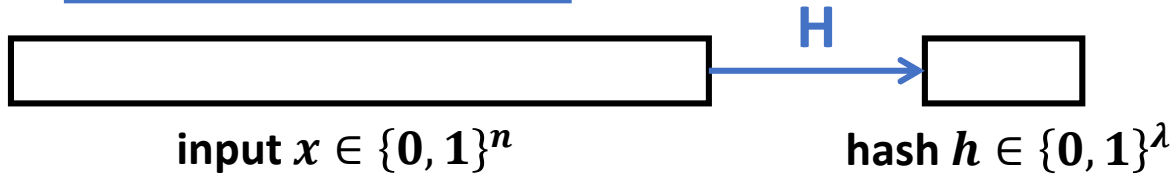


Alice



Bob

The Hash Function:



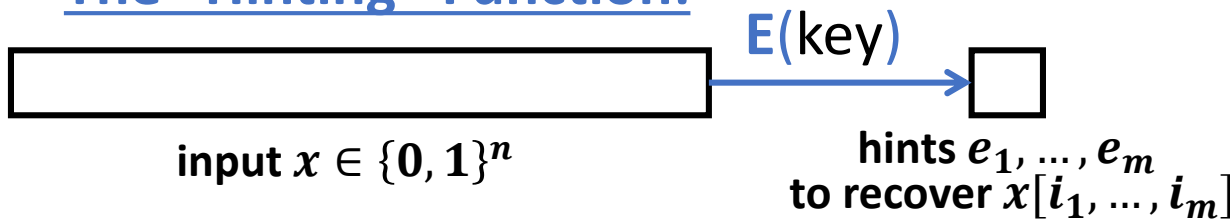
h

Input Privacy: h hides x .

h

“I want to learn $x[i_1, \dots, i_m]$, give me some hints”

The “Hinting” Function:



$\text{key}_1, \dots, \text{key}_m$

Key Generation:

$(\text{key}_j, \text{trapdoor}_j) \leftarrow \mathbf{G}(i_j)$

Index Privacy: key hides i .

Decoding:

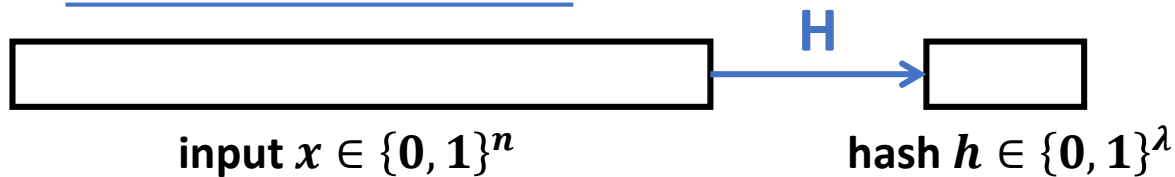
$x[i_j] \leftarrow \mathbf{D}(\text{trapdoor}_j, h, e_j)$

Efficiency: small hints, i.e. **high rate**

Defining Trapdoor Hash

TDH = (H,G,E,D)

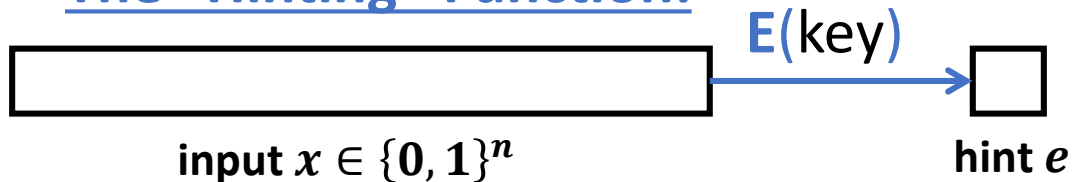
The Hash Function:



Key Generation:

$(\text{key}, \text{trapdoor}) \leftarrow \mathbf{G}(i)$

The “Hinting” Function:



Decoding:

$x[i] \leftarrow \mathbf{D}(\text{trapdoor}, h, e)$

Input Privacy: h hides x .

Index Privacy: key hides i .

Rate:

$$\frac{1}{|e|}$$

Optimally, **rate = 1**.

Main technical contribution:

Rate-1 TDH from DDH, QR, LWE, DCR

We also consider TDH for general functions of x .

Trapdoor Hash from DDH

Similar technique used to construct **IBE [DG17]**, **laconic OT [CDGGMP17]** and **Trapdoor Functions [GH18,GGH19]**.

- **Multiplicative abelian group \mathbb{G} of prime order p , with a public generator $g \in \mathbb{G}$.**

For all $g_1, g_2 \in \mathbb{G}$,

$$g_1 \cdot g_2 = g_2 \cdot g_1 \in \mathbb{G}$$

- **The DDH (Decisional Diffie-Hellman) assumption**

For **uniform** $a, b, c \in \mathbb{Z}_p$ and an element $g \in \mathbb{G}$,

$$(g^a, g^b, g^{ab}) \equiv (g^a, g^b, g^c)$$

Trapdoor Hash from DDH



Alice

public parameters

$2 \times n$ uniform group elements

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix} \in \mathbb{G}^{2 \times n}$$



Bob



input $x \in \{0, 1\}^n$

Trapdoor Hash from DDH



Alice

public parameters

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$



Bob

Hash Function:

$$H(x) = \prod_j g_{j,x[j]}$$

$$h = H(x)$$



$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$

1	0	0	...	1
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H

$$h \in \mathbb{G}$$

input $x \in \{0, 1\}^n$

Input Privacy: statistical*.

Trapdoor Hash from DDH



Alice

public parameters

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$



Bob

Hash Function:

$$H(x) = \prod_j g_{j,x[j]}$$

$$h = H(x)$$



Key Generation:

“I want to learn $x[i]$ ”

$$key = \begin{pmatrix} \tilde{g}_{1,0} & \tilde{g}_{2,0} & \tilde{g}_{3,0} & \dots & \tilde{g}_{n,0} \\ \tilde{g}_{1,1} & \tilde{g}_{2,1} & \tilde{g}_{3,1} & \dots & \tilde{g}_{n,1} \end{pmatrix}$$



trapdoor: uniform $t \in \mathbb{Z}_p$

$$key = \begin{pmatrix} g_{1,0}^t & \dots & g_{i,0}^t & \dots & g_{n,0}^t \\ g_{1,1}^t & \dots & g_{i,1}^t \cdot g & \dots & g_{n,1}^t \end{pmatrix}$$

Index Privacy: assuming DDH,

$$\begin{pmatrix} \tilde{g}_{1,0} & \tilde{g}_{2,0} & \tilde{g}_{3,0} & \dots & \tilde{g}_{n,0} \\ \tilde{g}_{1,1} & \tilde{g}_{2,1} & \tilde{g}_{3,1} & \dots & \tilde{g}_{n,1} \end{pmatrix} \equiv \text{uniform matrix in } \mathbb{G}^{2 \times n}$$

Trapdoor Hash from DDH



Alice

public parameters

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$



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trapdoor: uniform $t \in \mathbb{Z}_p$

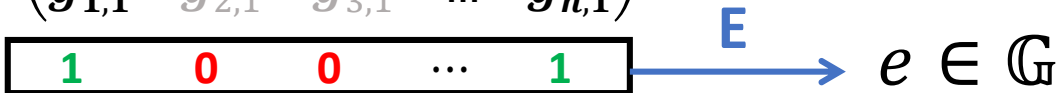
$$key = \begin{pmatrix} g_{1,0}^t & \dots & g_{i,0}^t & \dots & g_{n,0}^t \\ g_{1,1}^t & \dots & g_{i,1}^t \cdot g & \dots & g_{n,1}^t \end{pmatrix}$$

Hinting:

$$E(key, x) = \prod_j \tilde{g}_{j,x[j]}$$

$$\begin{pmatrix} \tilde{g}_{1,0} & \tilde{g}_{2,0} & \tilde{g}_{3,0} & \dots & \tilde{g}_{n,0} \\ \tilde{g}_{1,1} & \tilde{g}_{2,1} & \tilde{g}_{3,1} & \dots & \tilde{g}_{n,1} \end{pmatrix}$$

$$e = E(key, x)$$



input $x \in \{0, 1\}^n$

Rate:

$$\frac{1}{|e|} = \frac{1}{\lambda}$$

Trapdoor Hash from DDH



Alice

public parameters

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$



Bob

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$$H(x) = \prod_j g_{j,x[j]}$$

$$h = H(x)$$



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trapdoor: uniform $t \in \mathbb{Z}_p$

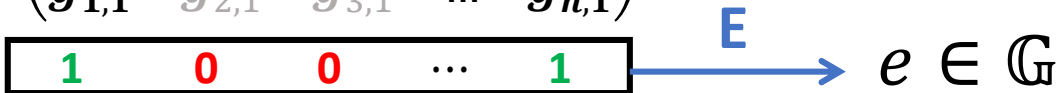
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$$e = E(key, x)$$



input $x \in \{0, 1\}^n$

How to recover $x[i]$ given

- The hash h
- The hint e
- The trapdoor t ? 22

Trapdoor Hash at The Bar: Decoding

Observation: given that

$$\begin{pmatrix} \tilde{g}_{1,0} & \tilde{g}_{2,0} & \tilde{g}_{3,0} & \dots & \tilde{g}_{n,0} \\ \tilde{g}_{1,1} & \tilde{g}_{2,1} & \tilde{g}_{3,1} & \dots & \tilde{g}_{n,1} \end{pmatrix} = \begin{pmatrix} g_{1,0}^t & \dots & g_{i,0}^t & \dots & g_{n,0}^t \\ g_{1,1}^t & \dots & g_{i,1}^t \cdot g & \dots & g_{n,1}^t \end{pmatrix}$$

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix} \xrightarrow{\mathbf{H}} h = \prod_j g_{j,x[j]}$$

input $x \in \{0, 1\}^n$

$$\begin{pmatrix} \tilde{g}_{1,0} & \tilde{g}_{2,0} & \tilde{g}_{3,0} & \dots & \tilde{g}_{n,0} \\ \tilde{g}_{1,1} & \tilde{g}_{2,1} & \tilde{g}_{3,1} & \dots & \tilde{g}_{n,1} \end{pmatrix} \xrightarrow{\mathbf{E}} e = \prod_j \tilde{g}_{j,x[j]} = \prod_j g_{j,x[j]}^t \cdot g^{x[i]} = h^t \cdot g^{x[i]}$$

input $x \in \{0, 1\}^n$

Trapdoor Hash from DDH



Alice

public parameters

$$\begin{pmatrix} g_{1,0} & g_{2,0} & g_{3,0} & \dots & g_{n,0} \\ g_{1,1} & g_{2,1} & g_{3,1} & \dots & g_{n,1} \end{pmatrix}$$



Bob

Hash Function:

$$H(x) = \prod_j g_{j,x[j]}$$

$$h = H(x)$$



Key Generation:

“I want to learn $x[i]$ ”

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trapdoor: uniform $t \in \mathbb{Z}_p$

$$key = \begin{pmatrix} g_{1,0}^t & \dots & g_{i,0}^t & \dots & g_{n,0}^t \\ g_{1,1}^t & \dots & g_{i,1}^t \cdot g & \dots & g_{n,1}^t \end{pmatrix}$$

Hinting:

$$E(key, x) = \prod_j \tilde{g}_{j,x[j]}$$

$$e = E(key, x)$$



Decoding: using hash h and **trapdoor** t

$$e = h^t \rightarrow x[i] = 0$$

$$e = h^t \cdot g \rightarrow x[i] = 1$$

Rate- $1/\lambda$ Trapdoor Hash

- For applications in **Setting II (Sublinear Secure RAM Computation)**:
Rate- $1/\lambda$ TDH **is sufficient***.
- For applications in **Setting I (Rate-1 OT)**:
We need **Rate-1 TDH**, i.e. TDH where the hint is a single bit.

Rate-1 Trapdoor Hash from DDH



Alice



Bob

Hinting:

Decoding: using hash h and trapdoor t

$$e = \prod_i \tilde{g}_{i,x[i]}$$

$x[i] = 0$ ↙

$x[i] = 1$ ↘

$h^t \in \mathbb{G}$ $h^t \cdot g \in \mathbb{G}$

$$e = E(\text{key}, x) \longrightarrow$$

$= h^t ?$ ↙

$x[i] = 0$

e

$= h^t \cdot g ?$ ↘

$x[i] = 1$

Rate-1 Trapdoor Hash from DDH



Alice



Bob

Hinting:

$$e = \prod_i \tilde{g}_{i,x[i]}$$

$x[i] = 0$ ↙ ↘ $x[i] = 1$

$h^t \in \mathbb{G}$ $h^t \cdot g \in \mathbb{G}$

Decoding: using hash h and trapdoor t

$$\xrightarrow{LSB(e)}$$

$LSB(e)$

$= LSB(h^t)?$ ↙ ↘ $= LSB(h^t g)?$

$x[i] = 0$ $x[i] = 1$

Attempt I:

Send **LSB** of e .

Fails when $LSB(h^t) = LSB(h^t g)$ is **equal**: happens with probability $1/2$.

Rate-1 Trapdoor Hash from DDH



Alice



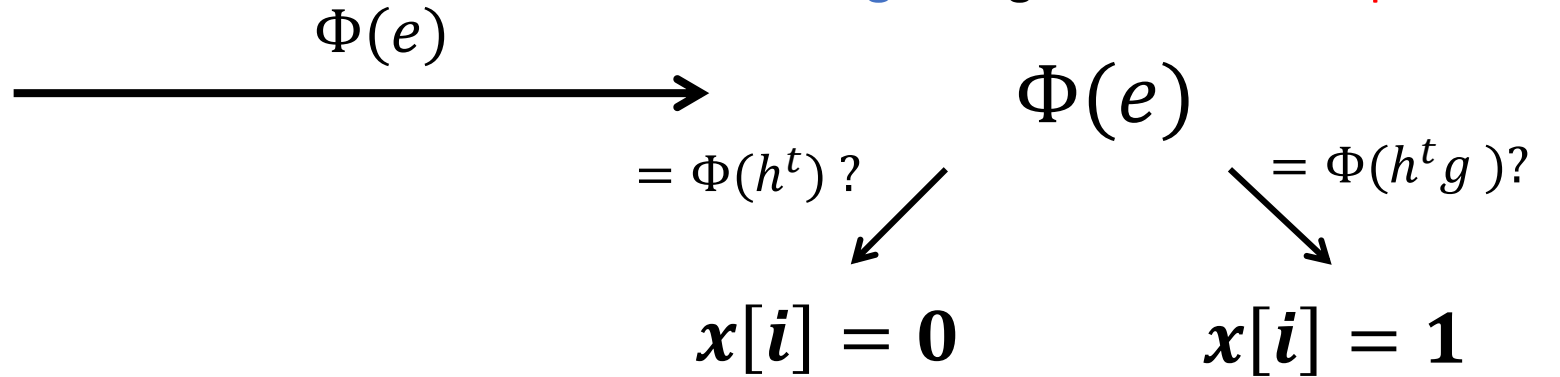
Bob

Hinting:

$$e = \prod_i \tilde{g}_{i,x[i]}$$

$x[i] = 0$ ↙ ↘ $x[i] = 1$
 $h^t \in \mathbb{G}$ $h^t \cdot g \in \mathbb{G}$

Decoding: using hash h and trapdoor t



Attempt 1:

Send **LSB** of e .

Fails when $LSB(h^t) = LSB(h^t g)$ is **equal**: happens with probability 1/2.

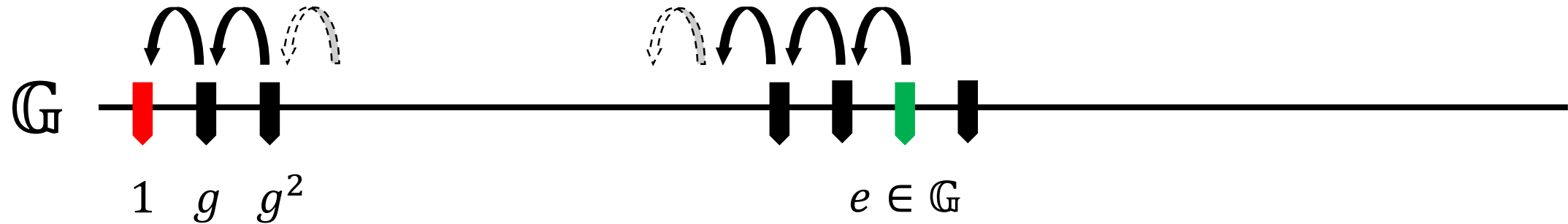
Goal: an encoding $\Phi: \mathbb{G} \rightarrow \{0,1\}$ such that with high probability

$$\Phi(h^t) \neq \Phi(h^t g)$$

Rate-1 Trapdoor Hash from DDH

Goal: an encoding $\Phi: \mathbb{G} \rightarrow \{0,1\}$ such that with high probability

$$\Phi(h^t) \neq \Phi(h^t g)$$



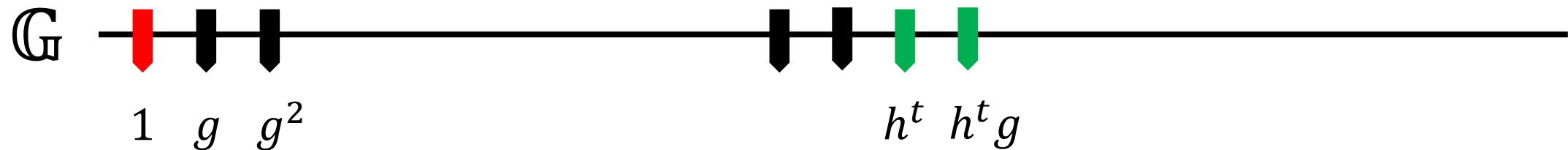
Attempt II:

$$\Phi(e) = \text{LSB of number of steps to reach 1}$$

Rate-1 Trapdoor Hash from DDH

Goal: an encoding $\Phi: \mathbb{G} \rightarrow \{0,1\}$ such that with high probability

$$\Phi(h^t) \neq \Phi(h^t g)$$



Attempt II:

$$\Phi(e) = \text{LSB of number of steps to reach 1}$$

Clearly, $\Phi(h^t) \neq \Phi(h^t g)$ for all $t \in \mathbb{Z}_p$.

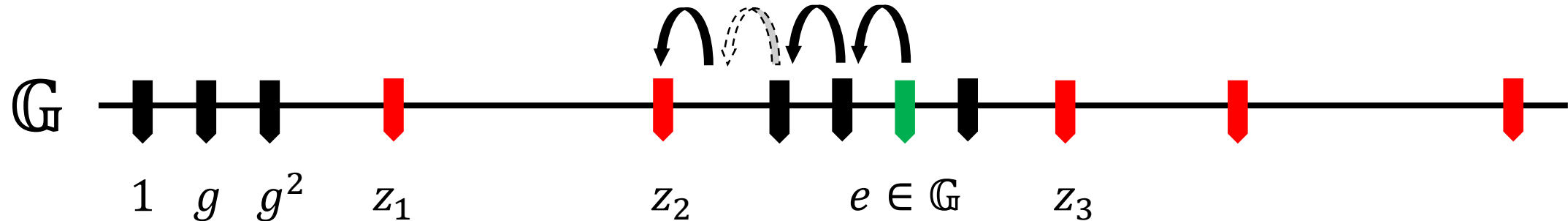
Problem: **not efficient** (this is DLOG).

Rate-1 Trapdoor Hash from DDH

Goal: an encoding $\Phi: \mathbb{G} \rightarrow \{0,1\}$ such that with high probability

$$\Phi(h^t) \neq \Phi(h^t g)$$

Idea: distributed discrete log [BGI16,DKK18]



Attempt III: using a PRF, define many random “**reference points**”: z s.t. $PRF(z) = 0$.

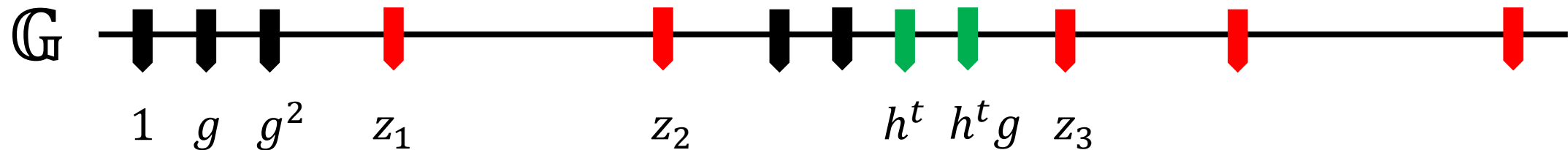
$\Phi(e) = \text{LSB of number of steps to reach the closest reference point}$

Rate-1 Trapdoor Hash from DDH

Goal: an encoding $\Phi: \mathbb{G} \rightarrow \{0,1\}$ such that with high probability

$$\Phi(h^t) \neq \Phi(h^t g)$$

Idea: distributed discrete log [BGI16,DKK18]



Attempt III: using a PRF, define many random “**reference points**”: z s.t. $PRF(z) = 0$.

$\Phi(e) = \text{LSB of number of steps to reach the closest reference point}$

If the reference points are **dense enough**, this is **efficient**.

If the reference points are **sparse enough**, $\Phi(h^t) \neq \Phi(h^t g)$ **with high probability**.

Conclusion and Further Research

We introduce **Trapdoor Hash**:

- Simple primitive with **constructions from various standard assumption**.
- Powerful primitive with **significant applications** in communication-efficient protocols.

Applications in **Setting I: Rate-Optimal Protocols for OT and Other Functions**:

- Constructions from new standard assumptions.
- More **general functionalities**?
- Constructions from **general assumptions**?

Applications in **Setting II: Sublinear Protocols for Secure RAM Computation**:

- First constructions from number-theoretic assumptions.
- Can we get constructions **as efficient as lattice-based schemes**?

Other applications of trapdoor hash or similar techniques?

Thanks for Listening!