

# GGHLite: More Efficient Multilinear Maps from Ideal Lattices

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# Our main result

Decrease size of public parameters from  $O(\lambda^5 \log \lambda)$  to  $O(\lambda \log^2 \lambda)$

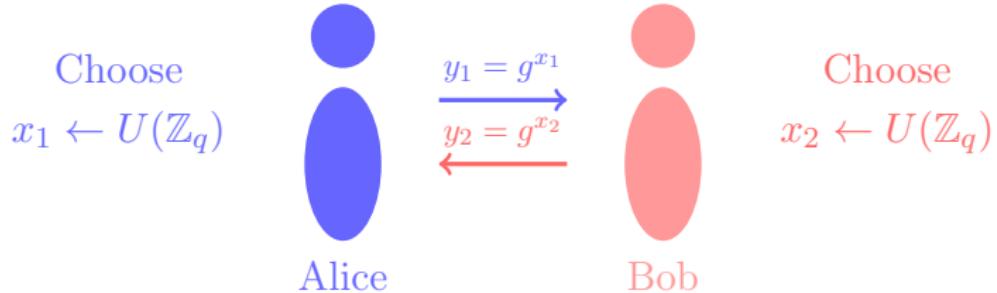
Lower size of parameters and finer security analysis

A more efficient cryptographic multilinear maps, obtained by formalizing, simplifying and improving the re-randomization process in the GGH construction.

For each encoding

Garg, Gentry and Halevi 2013

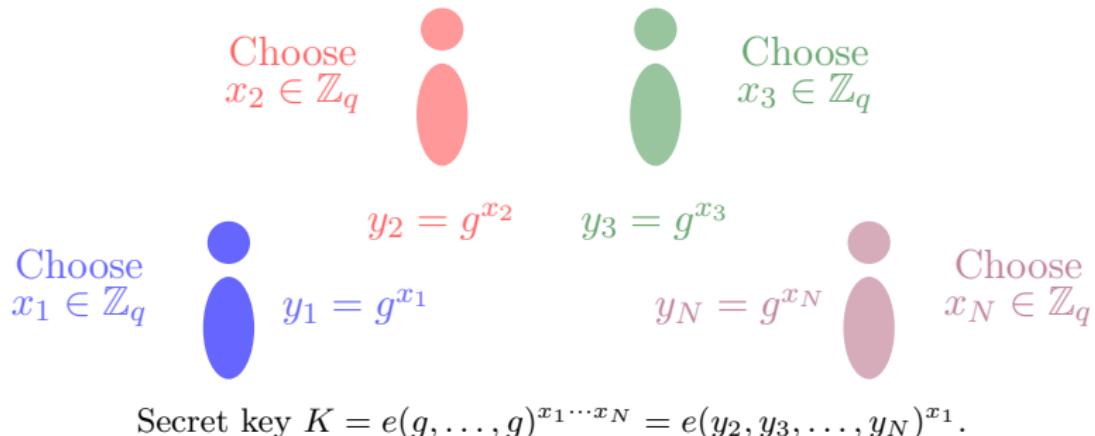
# Diffie-Hellman Key Exchange (1976)



$$\text{Agreed secret key: } K = g^{x_1 x_2} = y_1^{x_2} = y_2^{x_1}$$

- **Security:** **Decisional Diffie-Hellman** problem,  
**DDH:** For  $x_1, x_2, x_3 \leftarrow U(\mathbb{Z}_q)$ , distinguish between  
 $(g^{x_1}, g^{x_2}, g^{x_1 x_2})$  and  $(g^{x_1}, g^{x_2}, g^{x_3})$ .

# Cryptographic Multilinear Maps – 21st Century variant



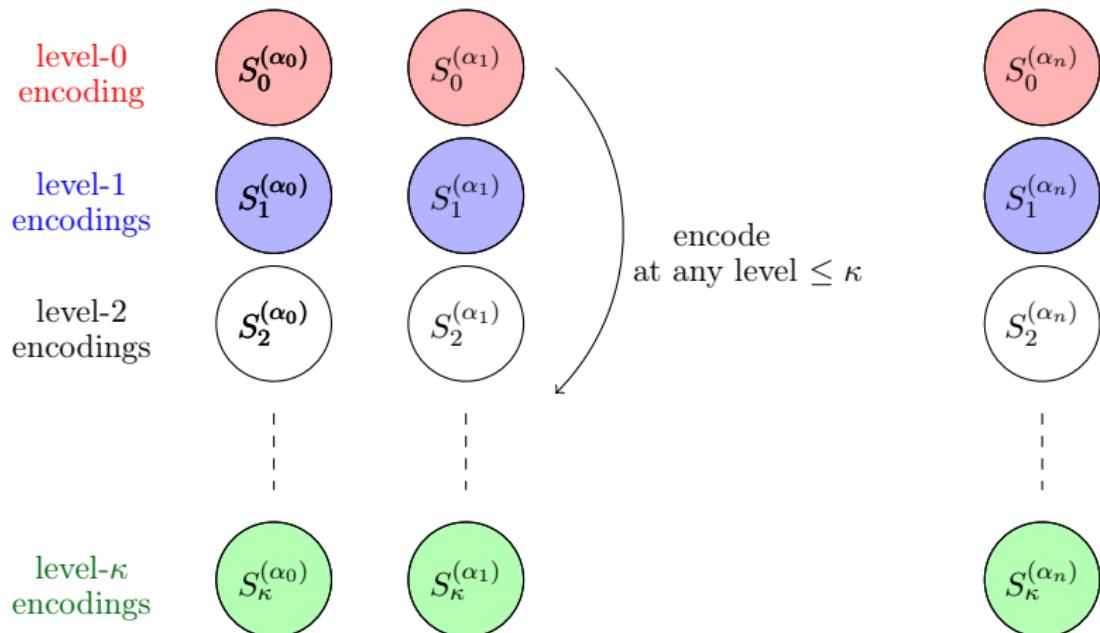
- **Security:** Hardness of **Multilinear Decisional DH** problem,  
**MDDH:** For  $x_1, \dots, x_N, x' \leftarrow U(\mathbb{Z}_q)$ , distinguish between  
 $(g^{x_1}, \dots, g^{x_N}, e(g, \dots, g)^{x_1 \cdots x_N})$  and  $(g^{x_1}, \dots, g^{x_N}, e(g, \dots, g)^{x'})$ .

# Cryptographic Multilinear Maps – History

- ▶ 2000: Applications for elliptic curves pairings ( $\kappa = 2$ )
  - ▶ 2000: 3-party non-interactive key agreement [Joux00],
  - ▶ 2000-2001: Identity-Based Encryption (IBE) ... [SakaiOhgishiKasahara00, BonehFranklin01],
- ▶ 2002: Applications for  $\kappa$ -linear maps [BonehSilverberg03]
  - ▶  $(\kappa + 1)$ -party non-interactive key agreement ...
- ▶ 2012: [GargGentryHalevi13]
  - ▶ First plausible realization for  $\kappa > 2$ , via ideal lattices,
  - ▶ Applications:
    - ▶ 2012-2013: Functional Encryption for arbitrary functions,
    - ▶ 2013: Program obfuscation notions for arbitrary functions.
- ▶ 2013: Variant over the integers [CoronLepointTibouchi13].
- ▶ 2014: **GGHLite** – More efficient variant of **GGH** (this talk).

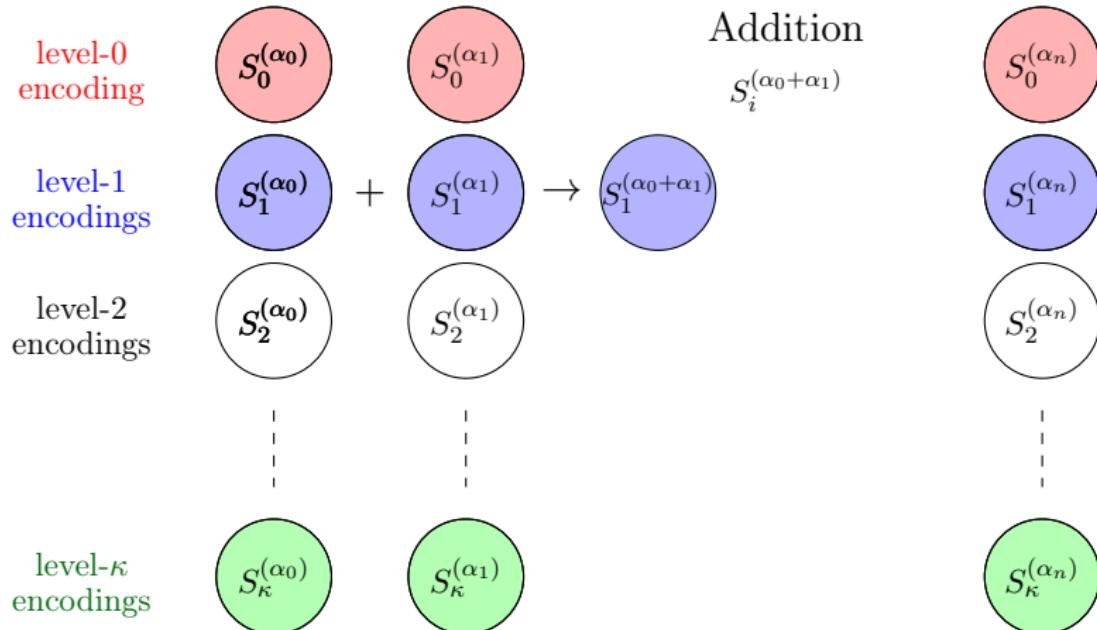
## $\kappa$ -Graded encoding scheme

A ring  $R_{\text{Plain}}$  of plaintext and a ring  $R_{\text{Enc}}$  of encodings,  
with a system of sets  $\mathcal{S} = \{S_i^{(\alpha)} \subseteq R_{\text{Enc}} : \alpha \in R_{\text{Plain}}, 0 \leq i \leq \kappa\}$ .



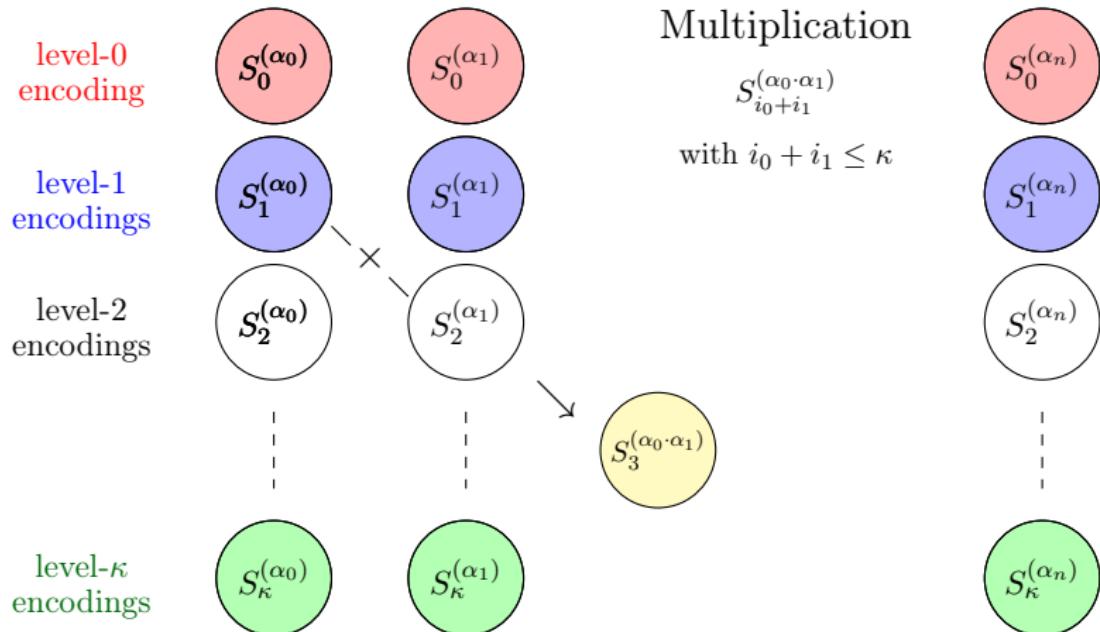
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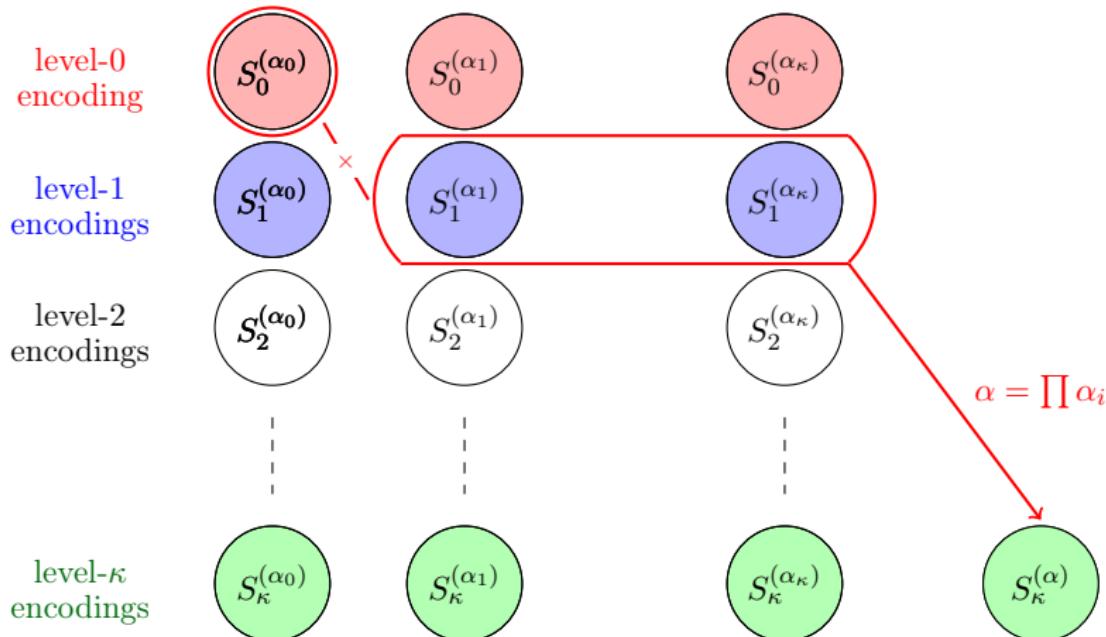
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## $\kappa$ -Graded encoding scheme – key exchange

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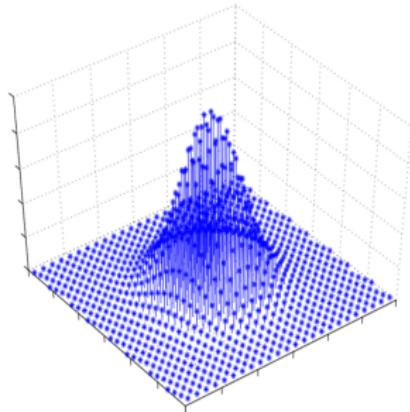
# Notations

- ▶ Polynomial Ring:  $R = \mathbb{Z}[x]/\langle x^n + 1 \rangle$  for  $n$  power of 2,
- ▶ Let  $q > 2$ . We let  $\textcolor{red}{R}_q = \mathbb{Z}_q[x]/\langle x^n + 1 \rangle$ ,
  - ▶ Arithmetic in  $R_q$  costs  $\tilde{O}(n \log q)$ .

## Discrete Gaussian on lattices

For a  $n$ -dimensional lattice  $\Lambda$ , a non-singular matrix  $\mathbf{S} \in \mathbb{R}^{n \times n}$ :

$$\forall x \in \Lambda : D_{\Lambda, \mathbf{S}, c}[x] \sim \exp(-\pi \|\mathbf{S}^{-1}(x - c)\|^2).$$



- ▶ small size (depending on  $\mathbf{S}$ ),
- ▶ sum is still Gaussian.

- **Instance generation**  $\text{InstGen}(1^\lambda, 1^\kappa)$ :
  - Sample  $g \leftarrow D_{R,\sigma}$  and  $z \leftarrow U(R_q)$ .
  - Sample a level-1 encoding of 1:  $\textcolor{orange}{y} = [\textcolor{blue}{a} \cdot z^{-1}]_q$  with  $\textcolor{blue}{a} \leftarrow D_{1+\langle g \rangle, \sigma'}$ .
  - For  $i \leq \kappa$ , sample  $m_r$  level- $i$  encodings of 0:  $(\textcolor{red}{x}_j^{(i)})_{j \leq m_r}$ .
  - Return public parameters  $\text{par} = (n, q, \textcolor{orange}{y}, \{\textcolor{red}{x}_j^{(i)}\}_{j \leq m_r, i \leq \kappa})$ .
- **Level- $k$  encoding**  $\text{enc}_k(e)$ : Given  $e \in R/\langle g \rangle$ :
  - ▶ Encode  $e$  at level  $k$ : Compute  $u' = [\textcolor{blue}{e} \cdot \textcolor{orange}{y}^k]_q$  ( $= [\textcolor{blue}{c}' / z^k]_q$ ).
  - ▶ Re-randomize: Sample  $\rho_j \leftarrow D_{\mathbb{Z}, \sigma^*}$  for  $j \leq m_r$  and return

$$u = [u' + \sum_{j=1}^{m_r} \rho_j \textcolor{red}{x}_j^{(k)}]_q = [(\textcolor{blue}{c}' + \sum_j \rho_j b_j^{(k)}) / z^k]_q$$

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Encoding of zero

# GGH

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Encoding of zero

Discrete Gaussian over  $\mathbb{Z}$

# Security

Which security?

- ▶ **Graded Decisional Diffie-Hellman – GDDH:**

Given  $\kappa + 1$  **level-1** encoding  $u_i$  of plaintexts  $e_i$ , distinguish between a **level- $\kappa$**  encoding of the product of the  $e_i$  and a **level- $\kappa$**  encoding of a random element.

To ensure security  $\Rightarrow$  **need randomization of the encodings**

- ▶ Without re-randomization,  $e$  can be efficiently recovered from  $u' = [e \cdot y]_q$  and  $y$  (by computing  $[u'y^{-1}]_q$ ).
- ▶ Re-randomization can prevent this attack.



With which parameters?

**This work:** understand the security of the **re-randomization** and propose efficient GGH variant achieving this security.

# GGHLite: Our contribution

We improve encoding re-randomization in GGH

- ▶  $m_r$  level-1 encodings of 0:  $\{x_j\}_{j \leq m_r}$ ,
- ▶ To randomize  $u' = [e \cdot y]_q$ , output  $u = [u' + \sum_j \rho_j b_j / z]_q$ ,
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By:

- ▶ Formalizing the security goal,
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- ▶ Formalizing the security goal,
  - ▶ Introduction of a **canonical version** of the problem.
- ▶ Decreasing the size of  $\sigma^*$ ,
  - ▶ Reduction from the canonical version to GCDH using **Rényi divergence** instead of the statistical distance.

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  - ▶ Introduction of a **canonical version** of the problem.
- ▶ Decreasing the size of  $\sigma^*$ ,
  - ▶ Reduction from the canonical version to GCDH using **Rényi divergence** instead of the statistical distance.
- ▶ Decreasing the number  $m_r$  of re-randomizers.
  - ▶ New **Leftover Hash Lemma**.

# GGH Lite: Formalizing Re-randomization Security

- ▶ **Informal requirement:** Prevent statistical correlation between re-randomized encoding and encoded element.
- ▶ **Formal requirement:** Breaking GCDH problem is as hard as breaking canonical GCDH problem.

$B$  has columns the  $b_j$ 's

We define:

GCDH:	canonical GCDH:
<p>Given <math>u_i = [(\textcolor{red}{c'_i} + \sum_{j=1}^{m_r} \rho_{j,i} \cdot \textcolor{red}{b_j})z^{-1}]_q</math> = <math>[\textcolor{blue}{c_i} z^{-1}]_q = \text{Enc}_1(e_i)</math></p> <p><math>\Rightarrow</math> compute level <math>\kappa</math> encoding of the product <math>\textcolor{blue}{c_1} \cdots \textcolor{blue}{c}_{\kappa+1}</math>.</p> <p><math>c_i \approx D_{\langle g \rangle + e_i, \sigma^* \textcolor{red}{B}^T, c'_i}</math> small centre <math>c'_i</math>.</p>	<p>Given <math>u_i = [\textcolor{blue}{c_i} z^{-1}]_q</math> with <math>\textcolor{blue}{c_i} \leftarrow D_{\langle g \rangle + e_i, \sigma^* \textcolor{red}{B}^T, 0}</math></p> <p><math>c_i \approx D_{\langle g \rangle + e_i, \sigma^* \textcolor{red}{B}^T, 0}</math> zero centre.</p>

# GGHLite Re-randomization Security: First Ingredient

Distribution of  $\textcolor{blue}{c}_i$  (with  $u_i = [\textcolor{blue}{c}_i/z]_q$ ):

<b>GCDH</b> $D_1 \approx \textcolor{blue}{D}_{\langle g \rangle + e_i, \sigma^* B^T, c'_i}$ small centre $c'_i$ .	<b>can-GCDH</b> $D_2 \approx \textcolor{blue}{D}_{\langle g \rangle + e_i, \sigma^* B^T, 0}$ zero centre.
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GGH security reduction based on **statistical distance** (SD):

$$\Delta(D_1, D_2) \stackrel{\text{def}}{=} \sum_x |D_1(x) - D_2(x)|,$$

Adversary  $\mathcal{A}_{\mathbf{GCDH}}$  with success  $\varepsilon \Rightarrow \mathcal{A}_{\text{canGCDH}}$  with success  $\varepsilon'$

$$\varepsilon' \geq \varepsilon - \Delta(D_1, D_2),$$

- ▶ To handle  $\varepsilon = 2^{-\lambda}$ , need  $\Delta(D_1, D_2) < 2^{-\lambda}$ .
- ▶ Consequently, need  $\frac{\sigma^*}{\|c'_i\|} = 2^{\Omega(\lambda)}$  (exponential drowning).

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GGHLite security reduction based on Rényi divergence (RD):

$$R(D_1 \| D_2) \stackrel{\text{def}}{=} \sum_x \frac{D_1^2(x)}{D_2(x)},$$

Adversary  $\mathcal{A}_{\mathbf{GCDH}}$  with success  $\varepsilon \Rightarrow \mathcal{A}_{\text{canGCDH}}$  with success  $\varepsilon'$

$$\varepsilon' \geq \frac{\varepsilon^2}{R(D_1 \| D_2)},$$

- ▶ Useful even if  $\varepsilon < R(D_1, D_2)^{-1}$  – use  $R(D_1 \| D_2) \leq \text{poly}(\lambda)$ .
- ▶ For  $R(D_1 \| D_2) \leq \text{poly}(\lambda)$ , can use  $\frac{\sigma^*}{\|c'_i\|} = O(\frac{1}{\log \lambda})$ .

# GGHLite: Second Main Ingredient

## New Leftover Hash Lemma

In GGH construction:

- ▶ Needs  $m_r = \Omega(n \log n)$  encodings of 0,
- ▶ Uses rational integer Gaussian randomizers ( $\rho_j \in \mathbb{Z}$ ),
- ▶ Uses a discrete Gaussian Leftover Hash Lemma (LHL) to show  $\sum_{j \leq m_r} \rho_j b_j$  distribution is close to a discrete Gaussian on  $\langle g \rangle$ .

GGHLite second ingredient:  $m_r = 2$  encodings of 0 suffice

- ▶ Uses Gaussian randomizers over full ring ( $\rho_j \in R$ ),
- ▶ New algebraic variant of discrete Gaussian LHL over  $R$ :  
 $\sum_{j \leq m_r} \rho_j b_j$  distribution is close to a discrete Gaussian on  $\langle g \rangle$ .

# GGHLite: Asymptotic Parameters

For  $\kappa$  level:

Parameter	GGHLite	GGH
$m_r$	2	$\Omega(n \log n)$
$\sigma^*$	$\tilde{O}(n^{5.5} \sqrt{\kappa})$	$\tilde{O}(2^\lambda \lambda n^{4.5} \kappa)$
$q$	$\tilde{O}((n^{10.5} \sqrt{\kappa})^{8\kappa})$	$\tilde{O}((2^\lambda \lambda^{1.5} n^{8.5} \kappa)^{8\kappa})$
$n$	$O(\kappa \lambda \log \lambda)$	$O(\kappa \lambda^2)$
$ \text{enc} $	$O(\kappa^2 \lambda \log^2 \lambda)$	$O(\kappa^2 \lambda^3)$
$ \text{par} $	$O(\kappa^3 \lambda \log^2 \lambda)$	$O(\kappa^4 \lambda^5 \log \lambda)$

# Adapting Applications of GGH to GGHLite

**Question:** How to adapt GGH applications  
to rely on **GCDH** rather than **GDDH**?

**Answer:** Replace  $K = v$  in original protocol by

$$K = H(v)$$

in modified protocol, where  $H(\cdot)$  is a cryptographic hash function.

If  $H(\cdot)$  is modelled as a **Random Oracle Model**, then security of modified protocol relies on GCDH – our GGHLite analysis applies.

# Conclusion

GGHLite: a more efficient variant of  
GGH graded encoding scheme.

- ▶ Reduction from **can-GCDH** to **GCDH** using Rényi divergence,
- ▶ New algebraic variant of discrete Gaussian LHL over  $R$ .

## Open Problems:

- ▶ Can our Rényi divergence analysis be applied to the **Decision** Graded Diffie Hellman problem?
- ▶ Understand the complexity of canonical GCDH problem – provable relation to standard lattice problems?
- ▶ Understand relation between GGH/GGHLite and more recent Jigsaw puzzle variants (obfuscation).
- ▶ Concrete computational / space efficiency of GGHLite based on best known attacks?