

TinyLabels: How to Compress Garbled Circuit Input Labels, Efficiently

Marian Dietz

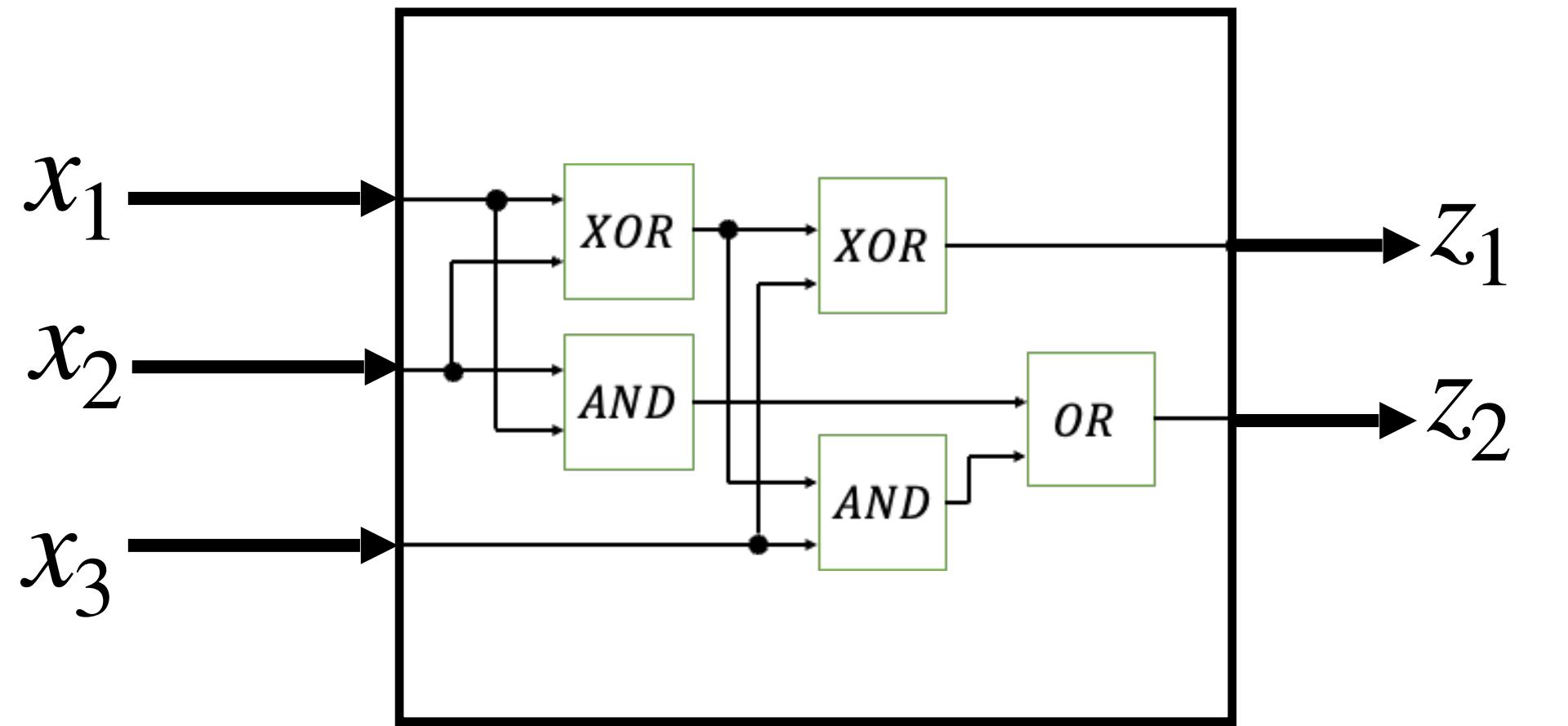
ETH Zurich

Hanjun Li

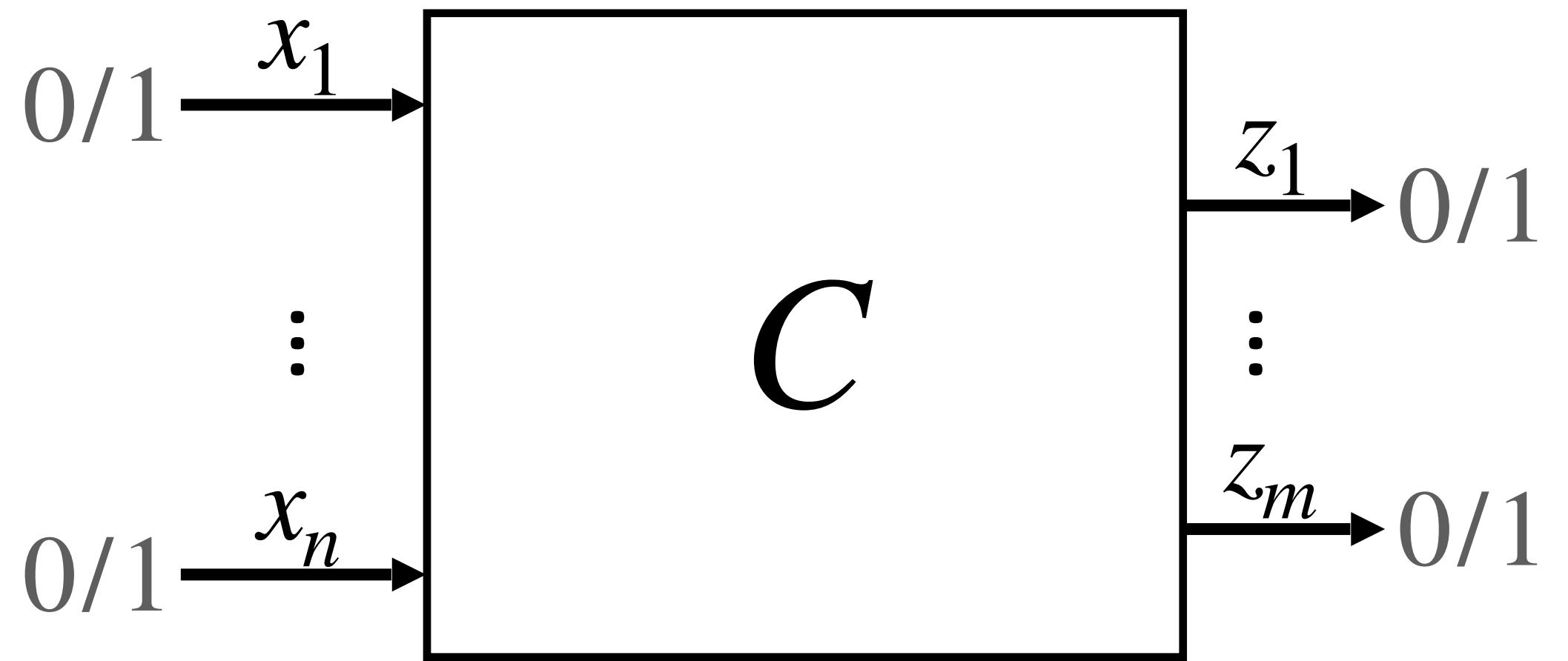
University of Washington

Huijia (Rachel) Lin

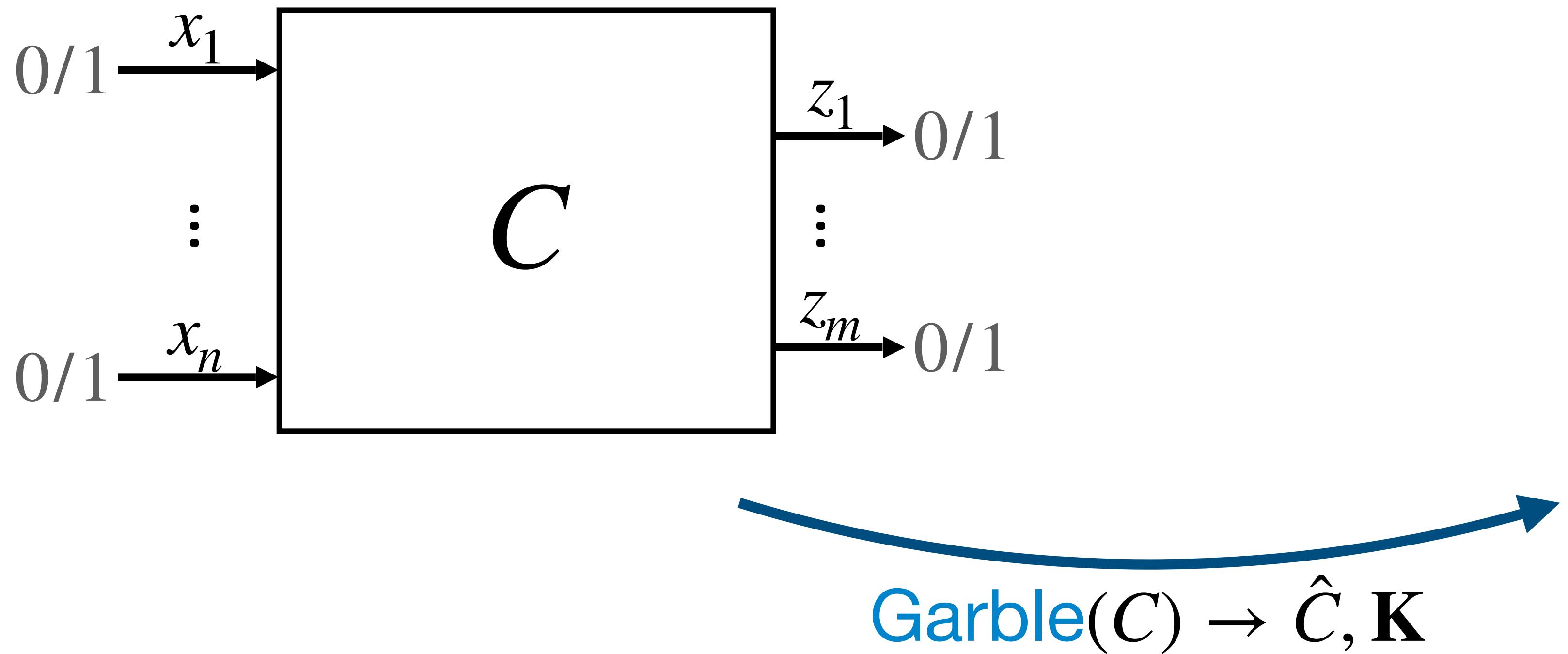
Garbling Scheme



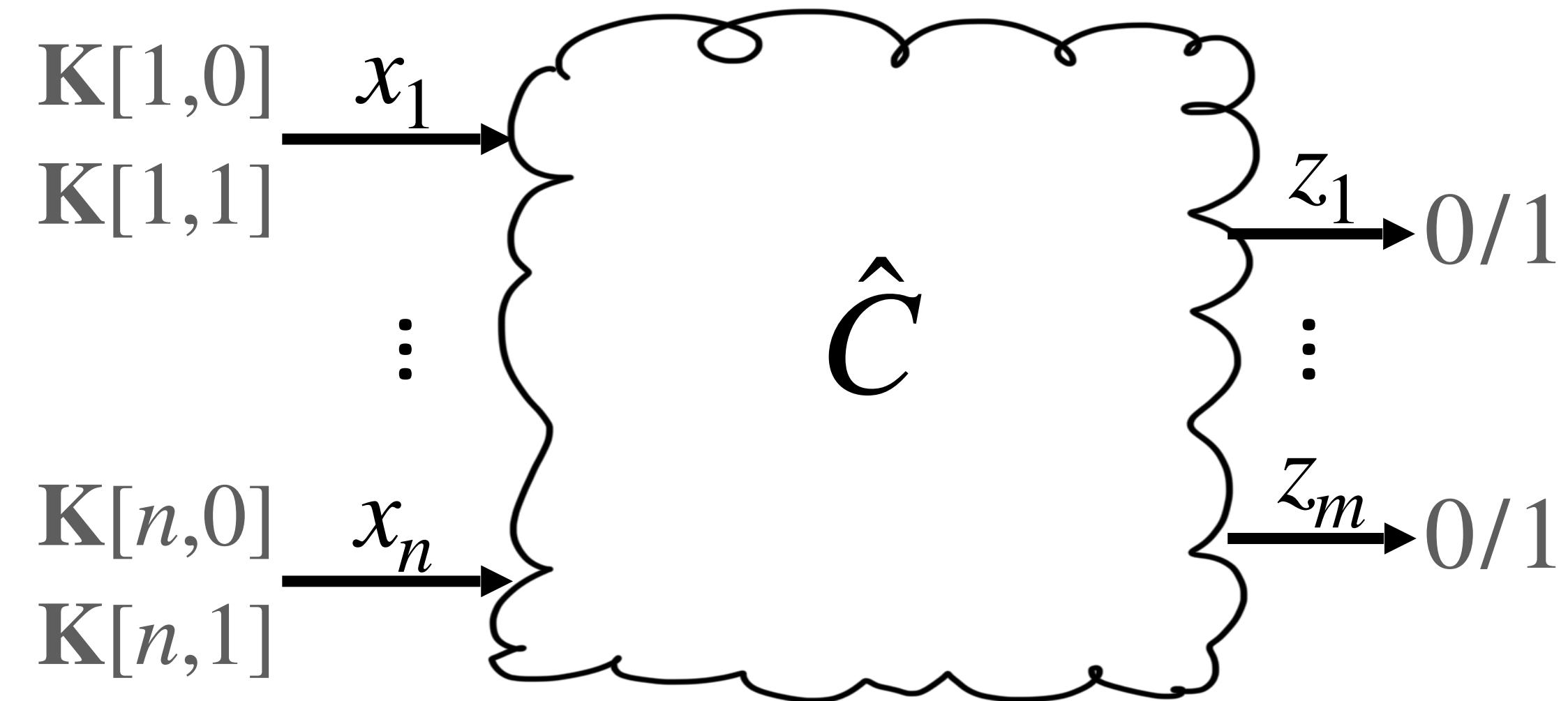
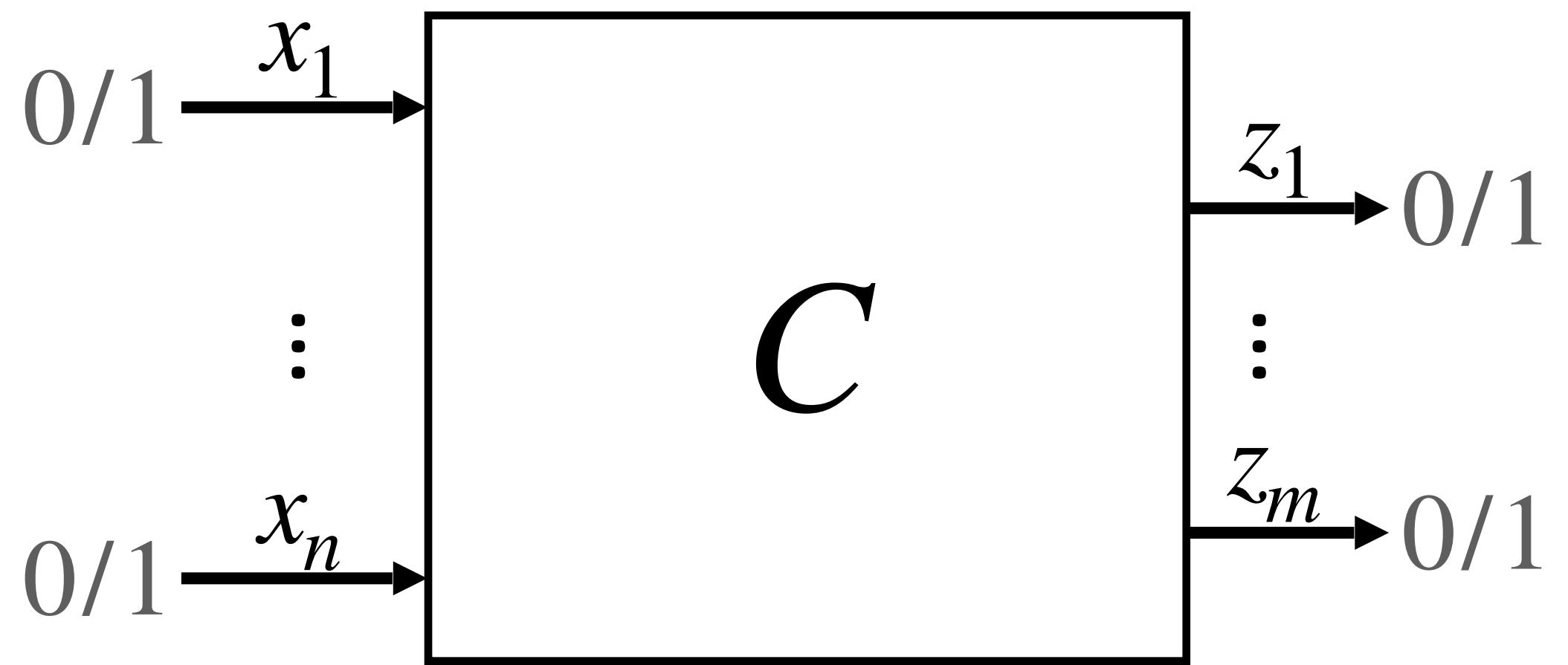
Garbling Scheme



Garbling Scheme

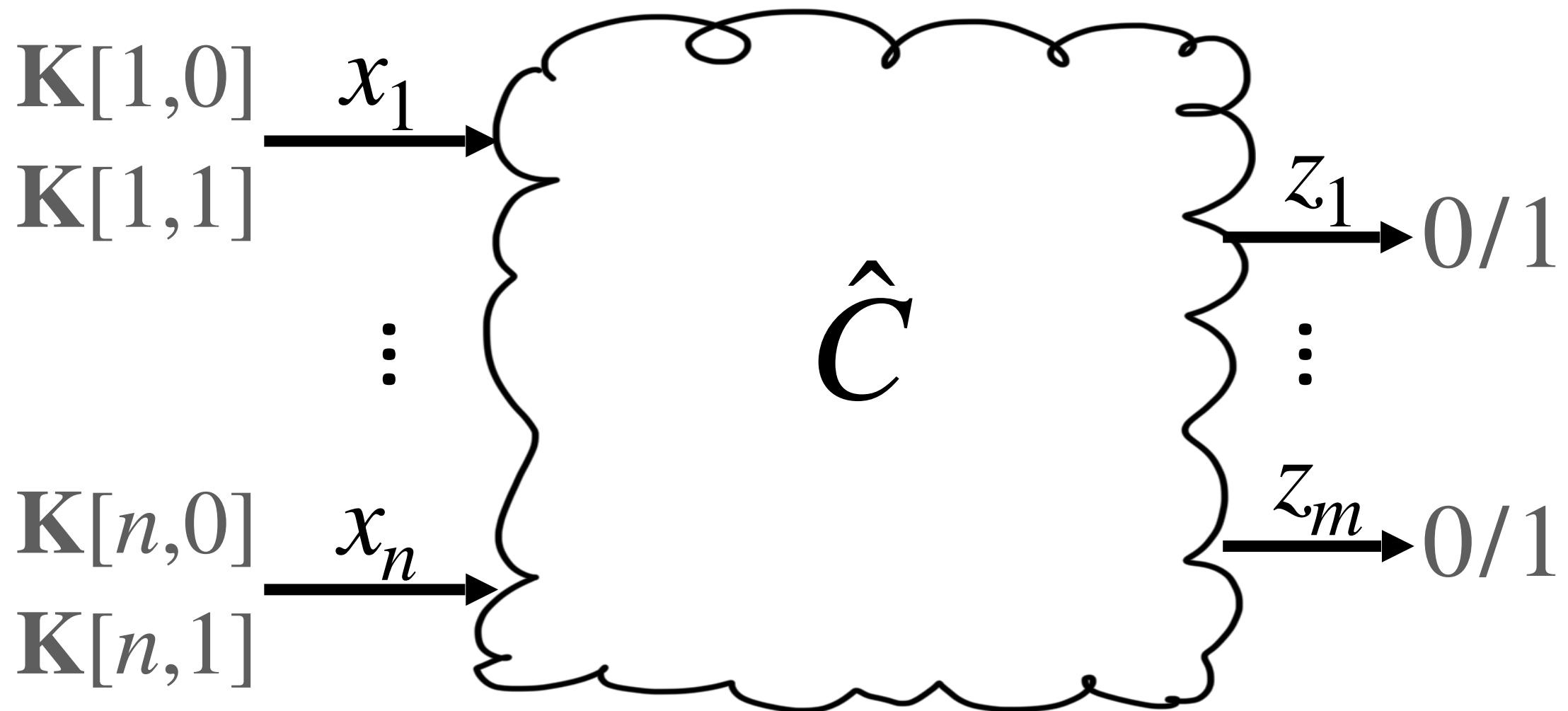
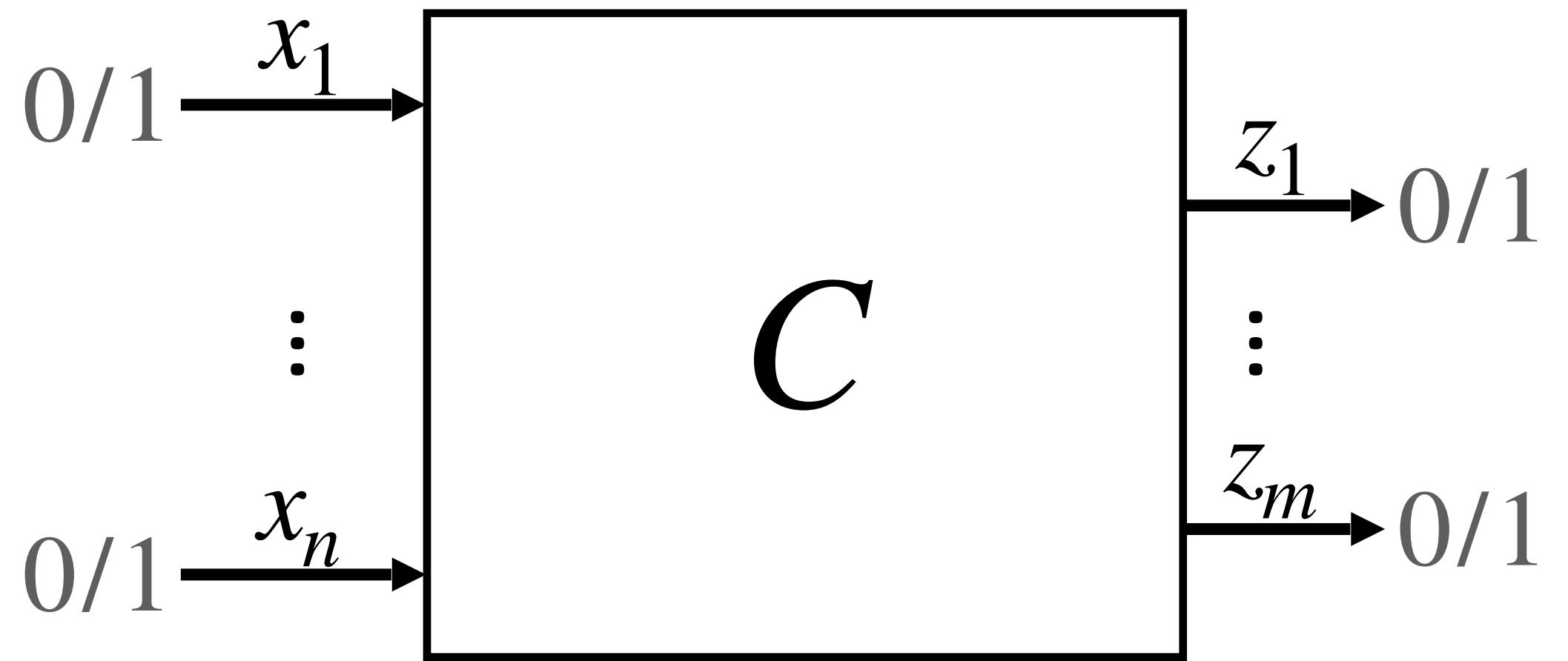


Garbling Scheme



Garble(C) $\rightarrow \hat{C}, \mathbf{K}$

Garbling Scheme

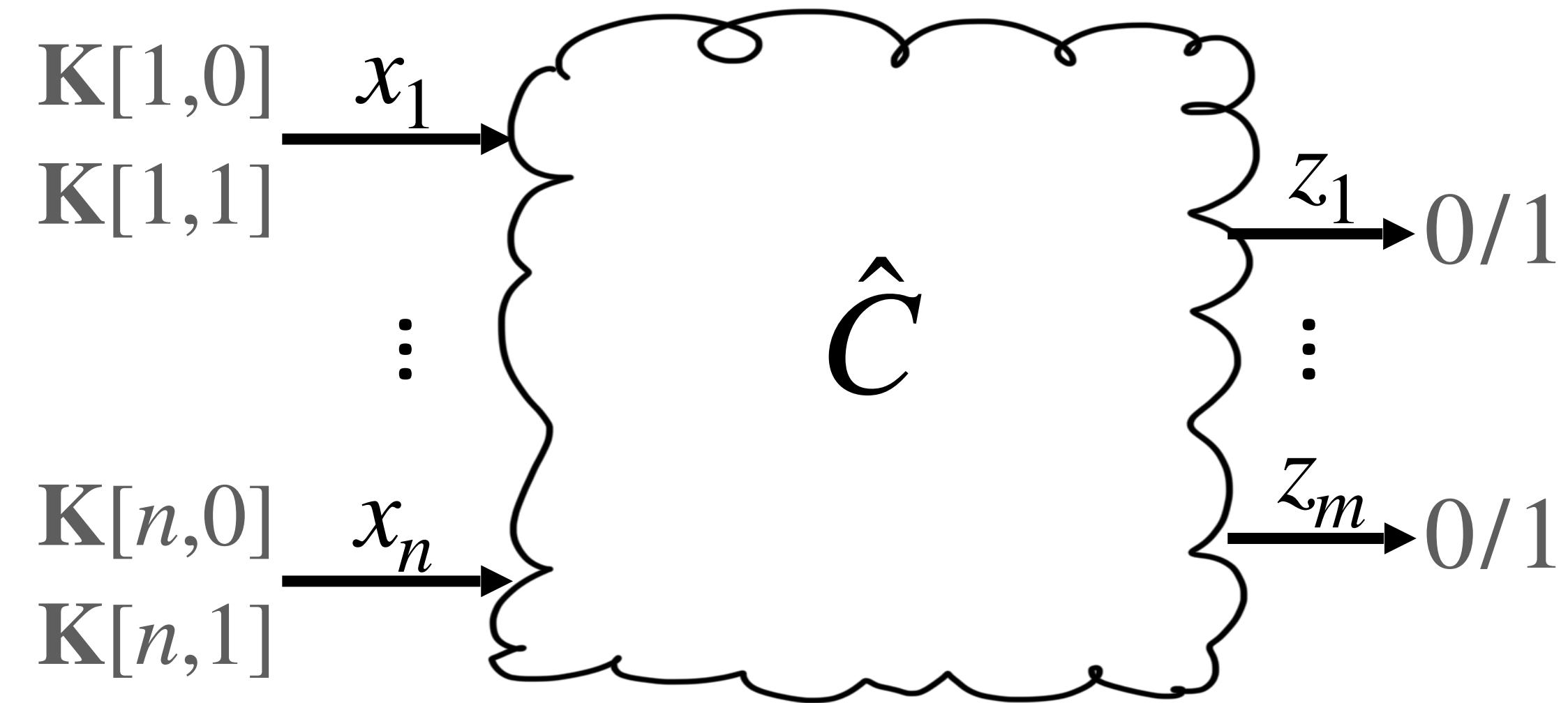
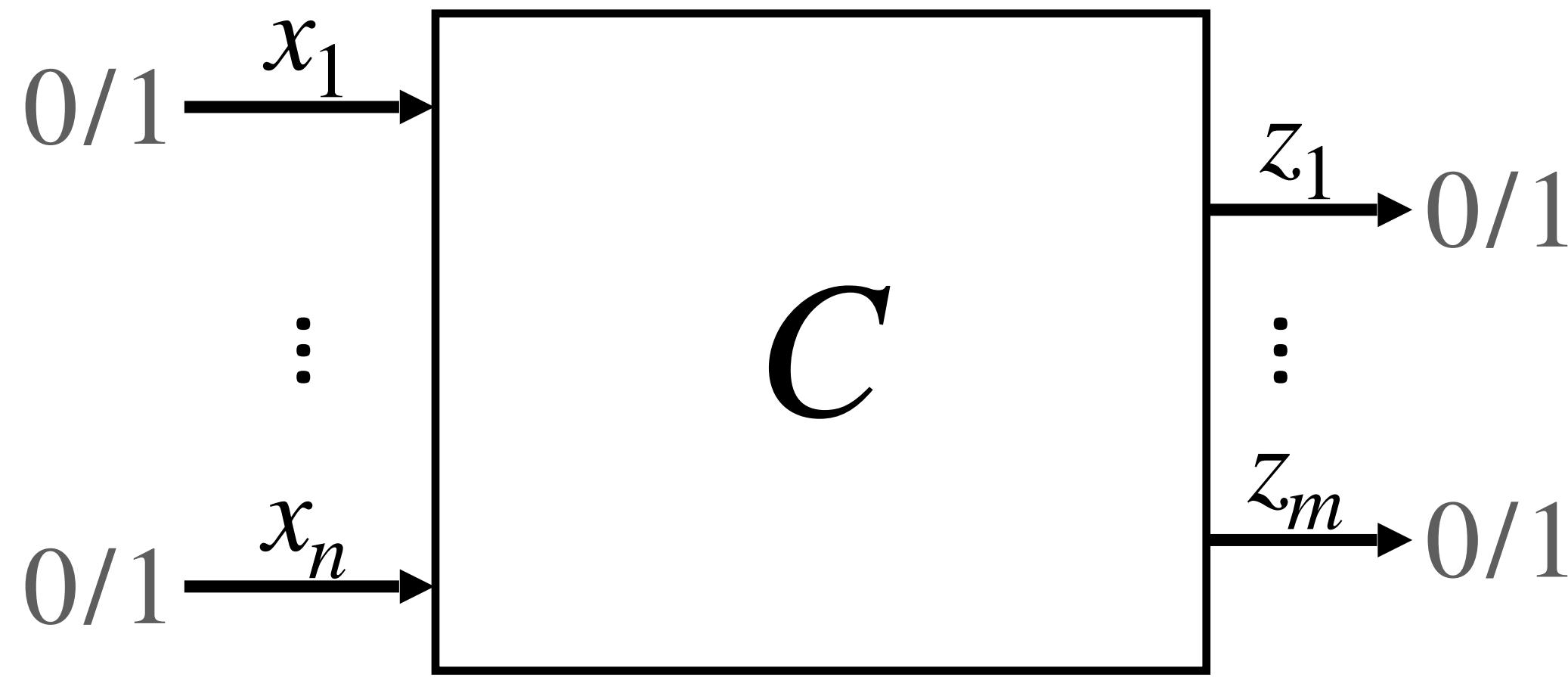


$\text{Garble}(C) \rightarrow \hat{C}, K$

$\text{Eval}(\hat{C}, K[x]) \rightarrow C(x)$

Input Labels: One key
 $K[i, x_i]$ per input bit

Garbling Scheme



$\text{Garble}(C) \rightarrow \hat{C}, \mathbf{K}$

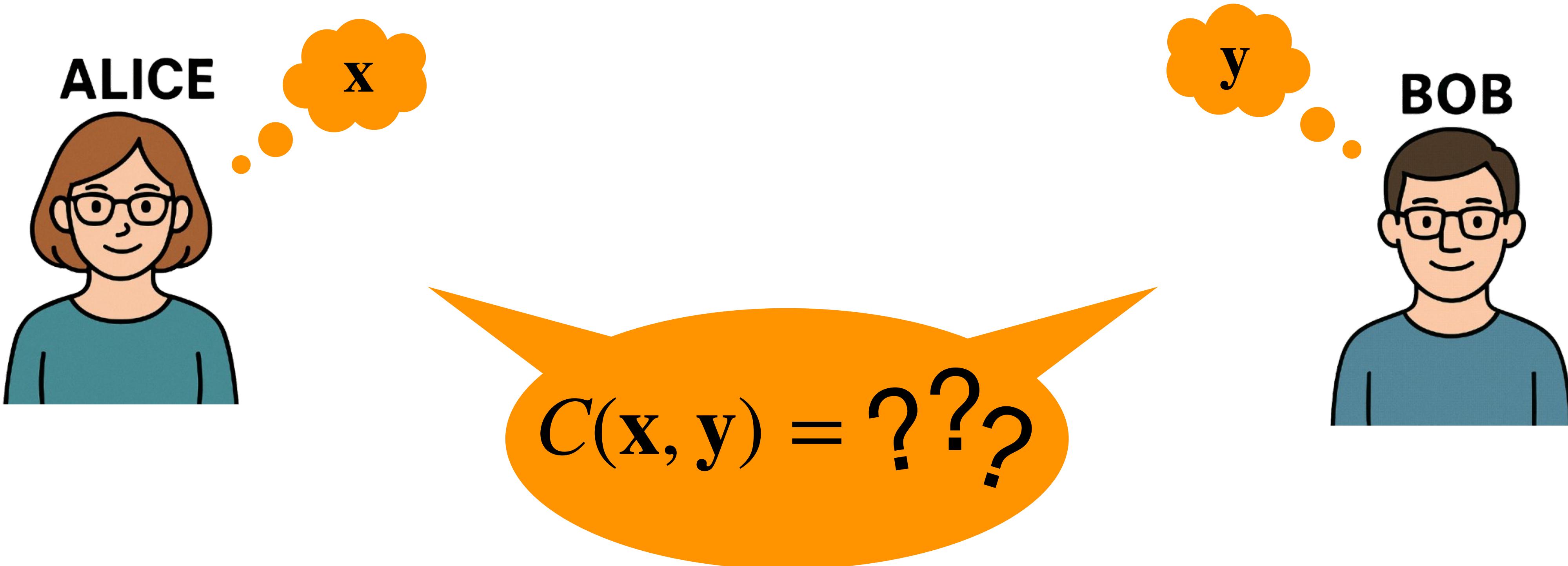
$\text{Eval}(\hat{C}, \mathbf{K}[x]) \rightarrow C(x)$

**Simulation
Security:**

$\text{Sim}(C(x)) \rightarrow \tilde{C}, L \approx \hat{C}, \mathbf{K}[x]$

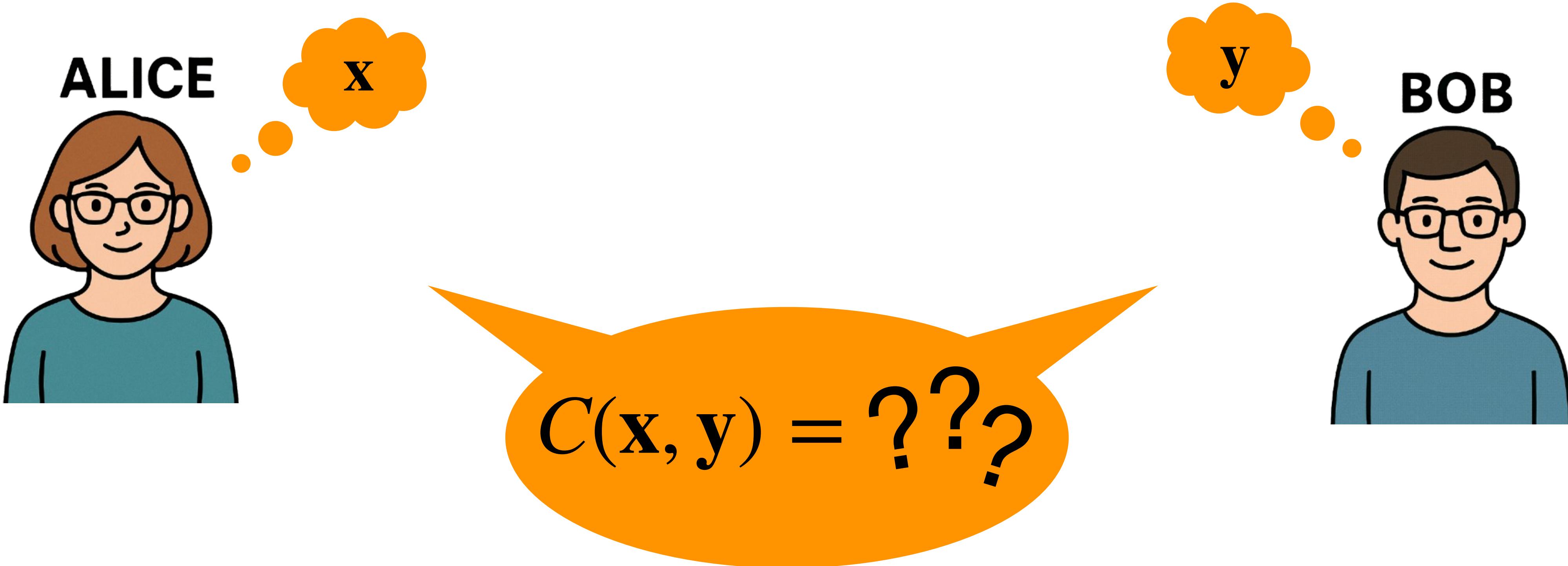
Garbling Schemes: Applications

- Two-party computation



Garbling Schemes: Applications

- Two-party computation



- Outsourcing computation on sensitive data

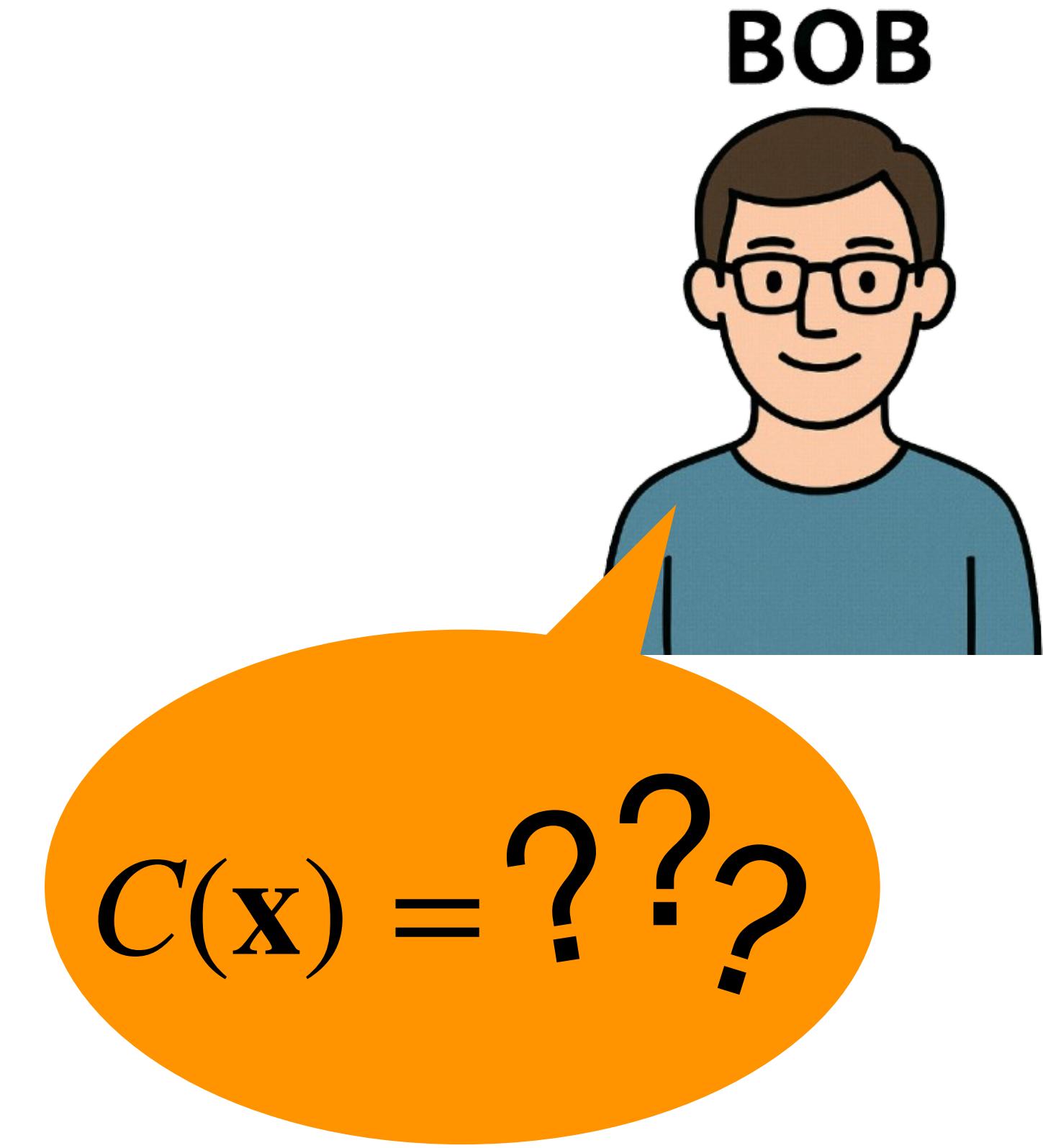
Application: Outsourcing computation

[AIKW13]



X

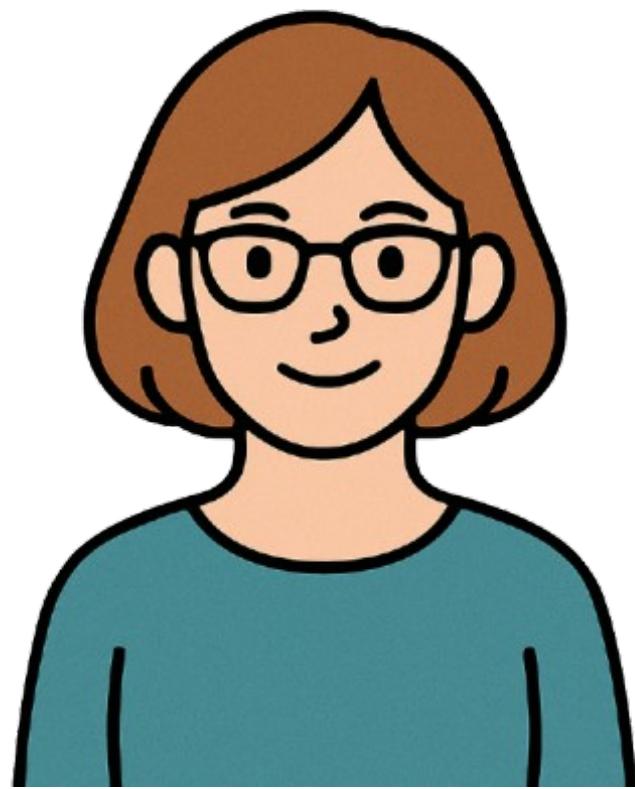
Alice has collected sensitive data \mathbf{x}
in a place with *limited resources*



Application: Outsourcing computation

[AIKW13]

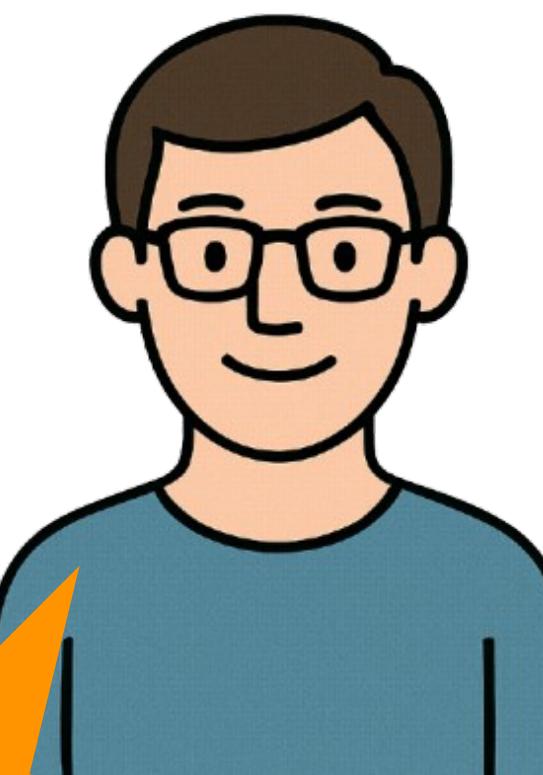
ALICE



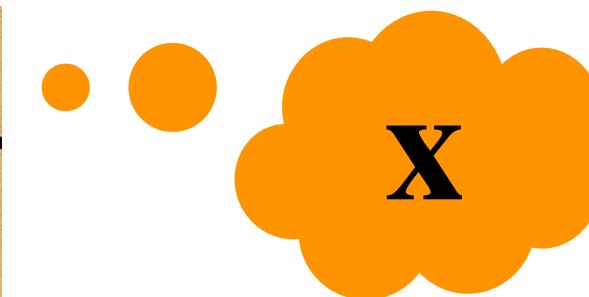
Offline (known C)

Alice has large resources

BOB



Online



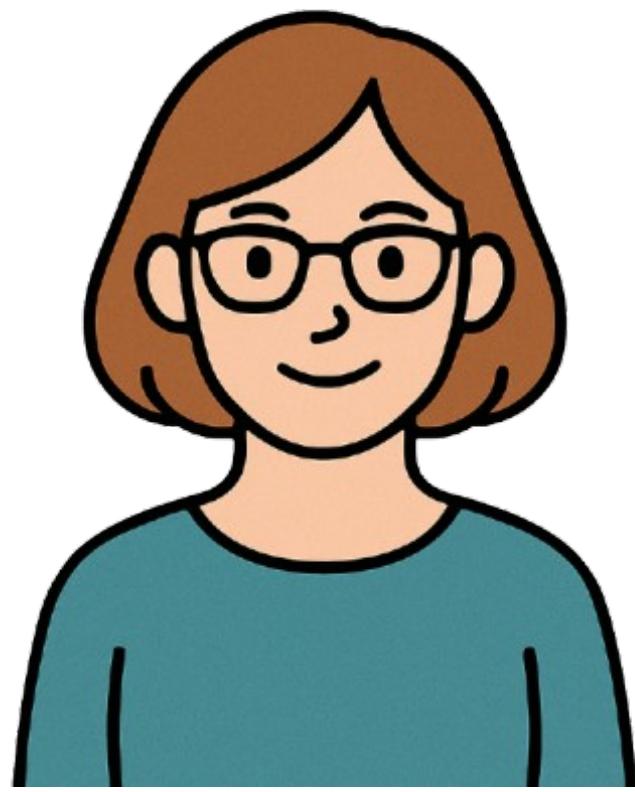
Alice has collected sensitive data \mathbf{x} in a place with *limited resources*

$C(\mathbf{x}) = ???$

Application: Outsourcing computation

[AIKW13]

ALICE

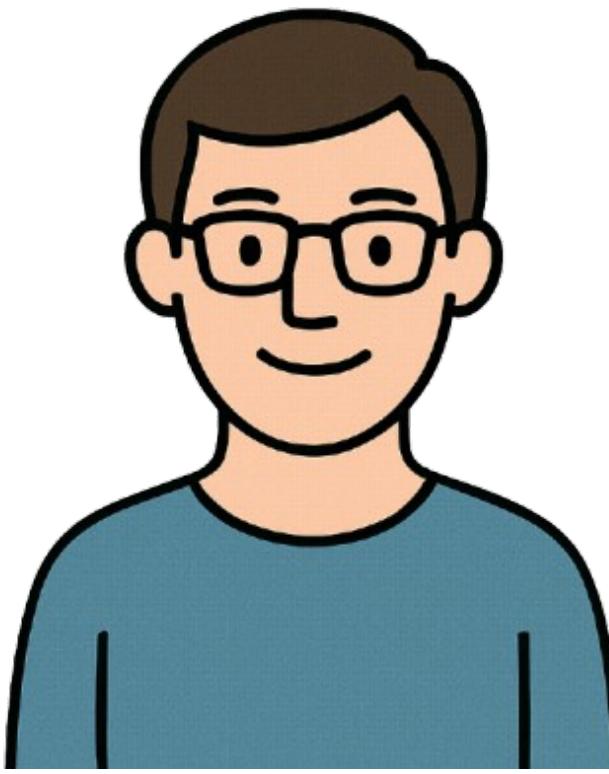


Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

\hat{C}

BOB



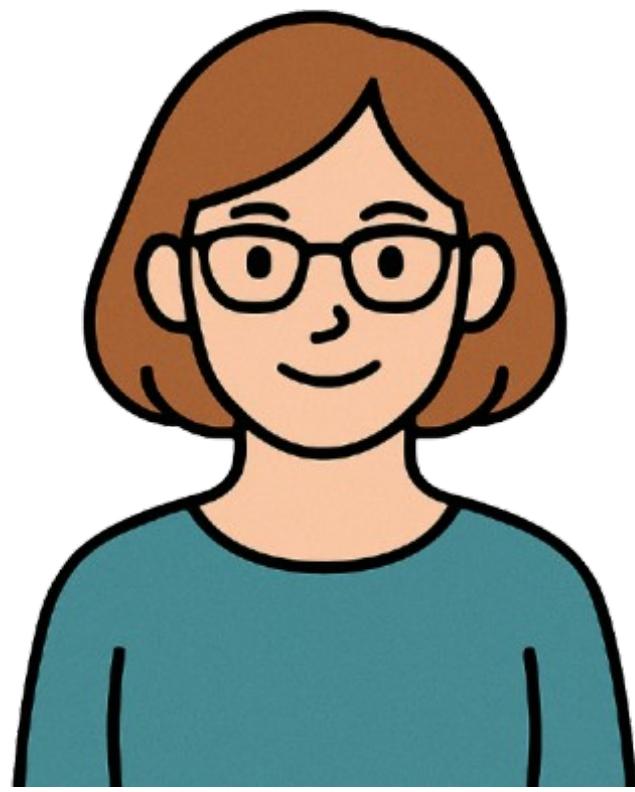
Online



Application: Outsourcing computation

[AIKW13]

ALICE

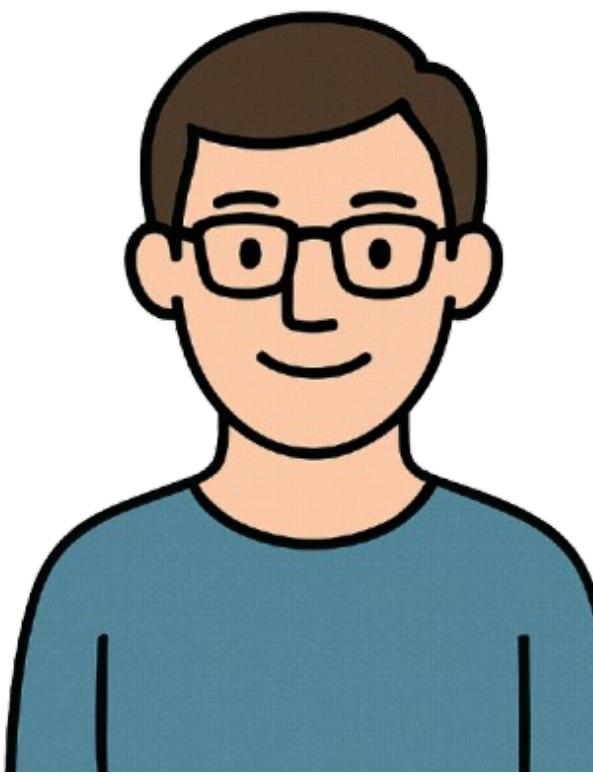


Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

\hat{C}

BOB



Online

$L = K[x]$

x

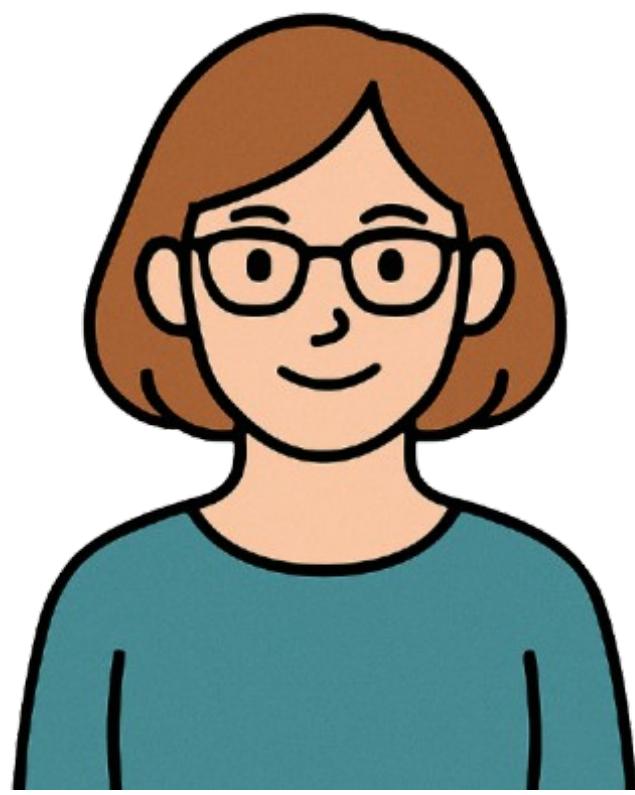


Eval(\hat{C}, L)
 $\rightarrow C(x)$

Application: Outsourcing computation

[AIKW13]

ALICE

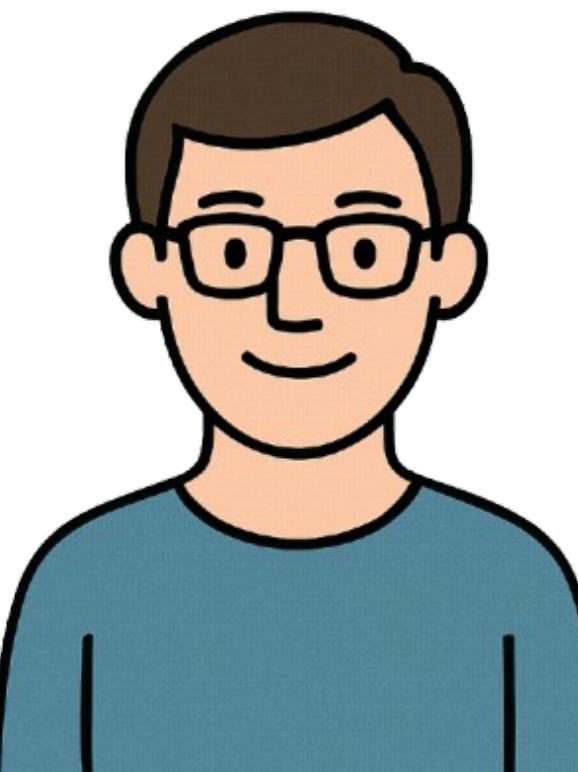


Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

\hat{C}

BOB



Online

$L = K[x]$

x

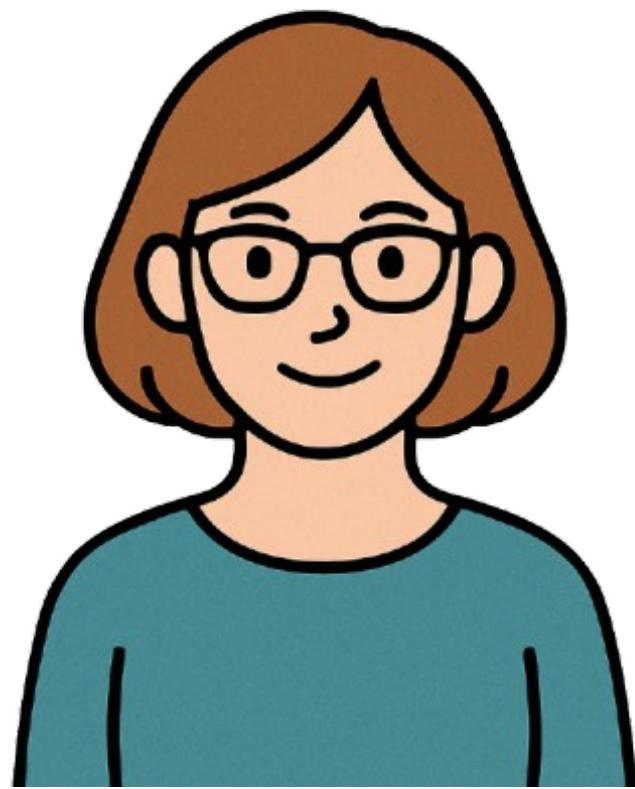
Size: $|x| \cdot \lambda$

Eval(\hat{C}, L)
 $\rightarrow C(x)$

Application: Outsourcing computation

[AIKW13]

ALICE

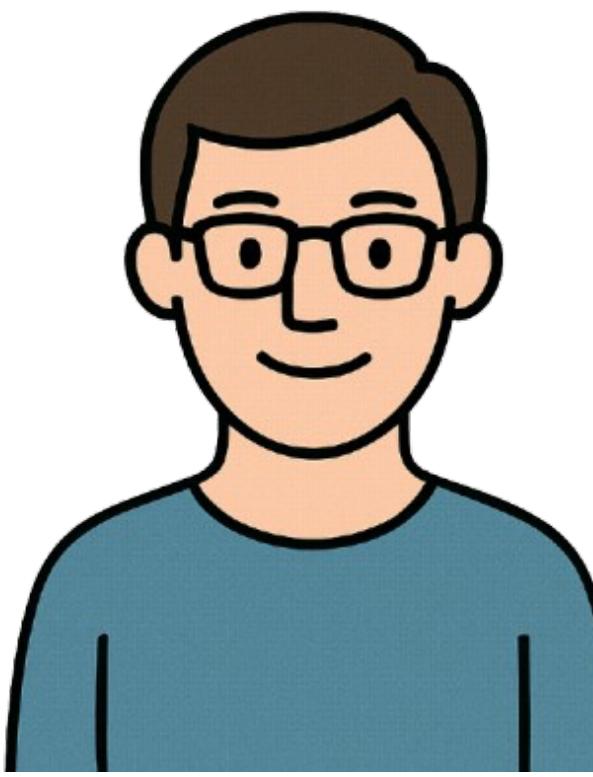


Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

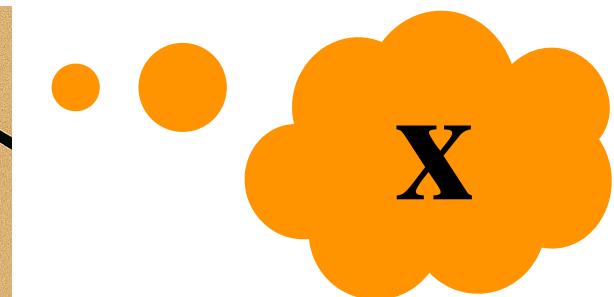
\hat{C}

BOB



Online

Compress($K[x]$, \bar{x})



Decompress $\rightarrow L$

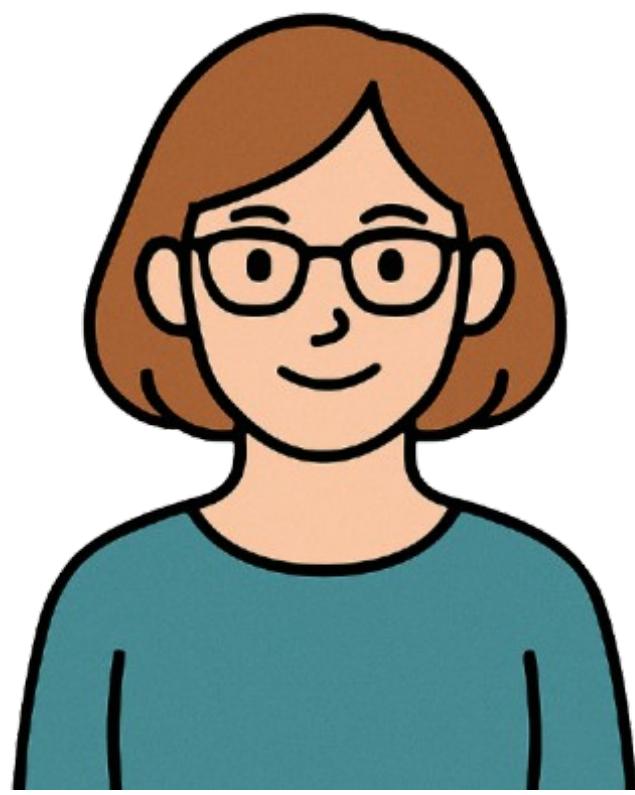
Eval(\hat{C}, L)
 $\rightarrow C(x)$



Application: Outsourcing computation

[AIKW13]

ALICE



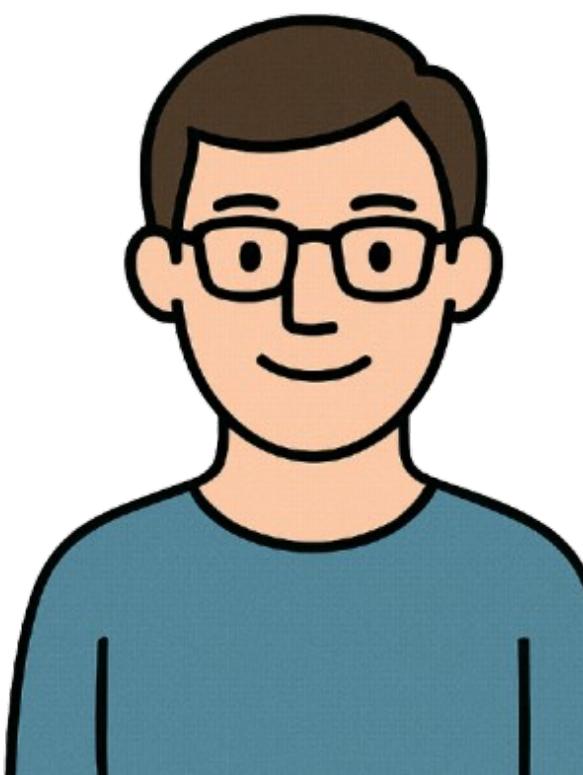
Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

\hat{C}

$CT(K)$

BOB



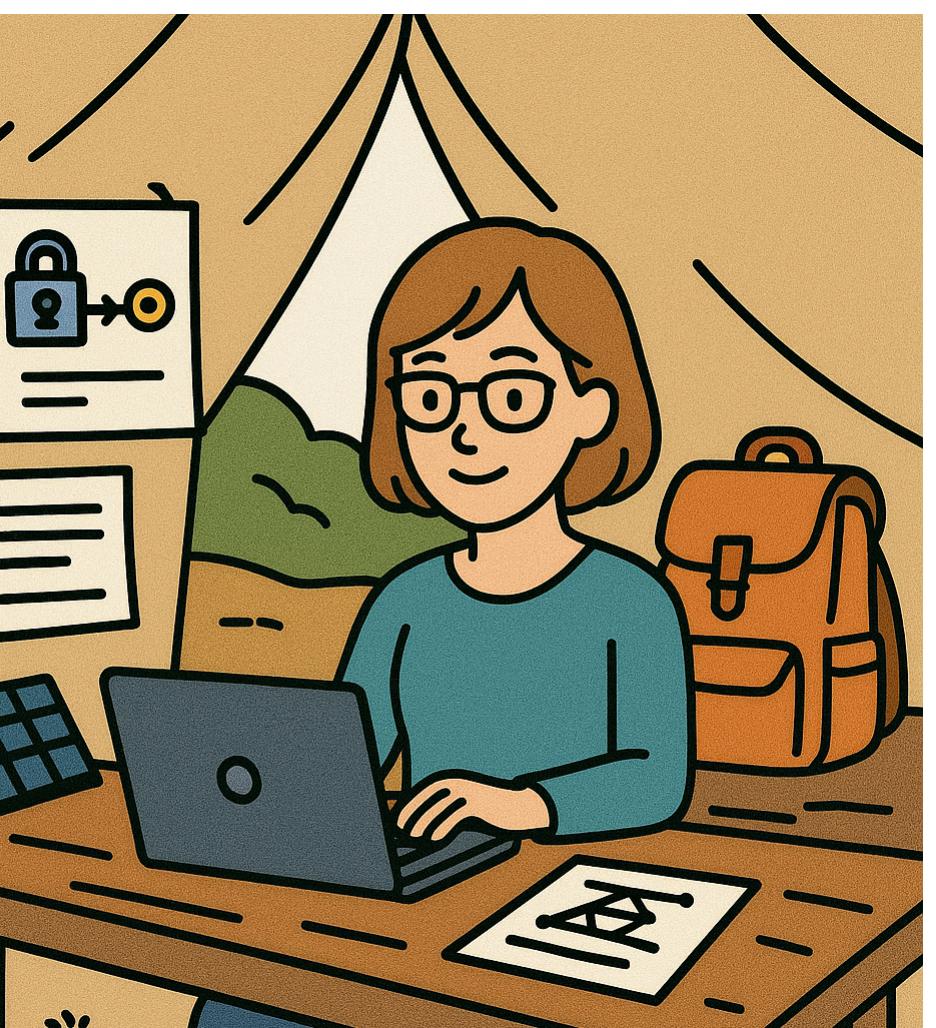
Online

Compress($K[x]$), \bar{x}

Decompress $\rightarrow L$

Eval(\hat{C}, L)
 $\rightarrow C(x)$

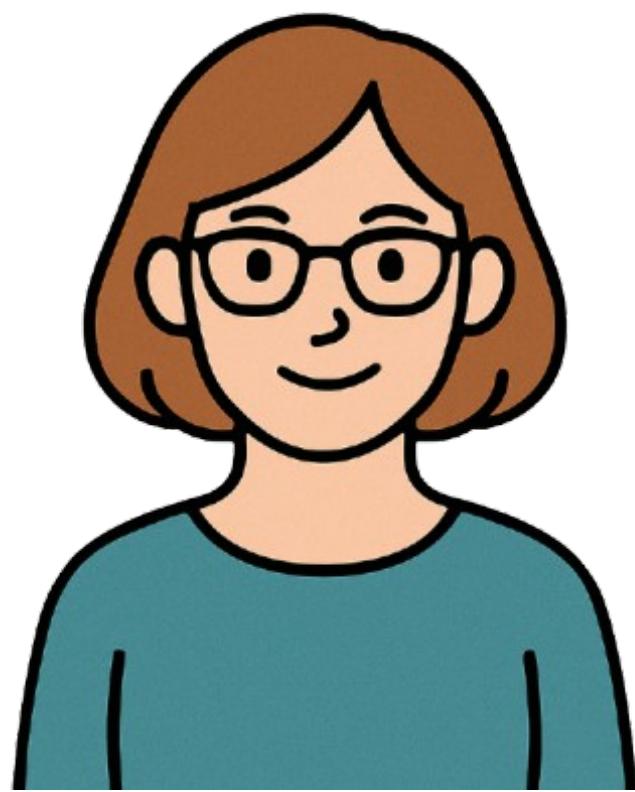
x



Application: Outsourcing computation

[AIKW13]

ALICE



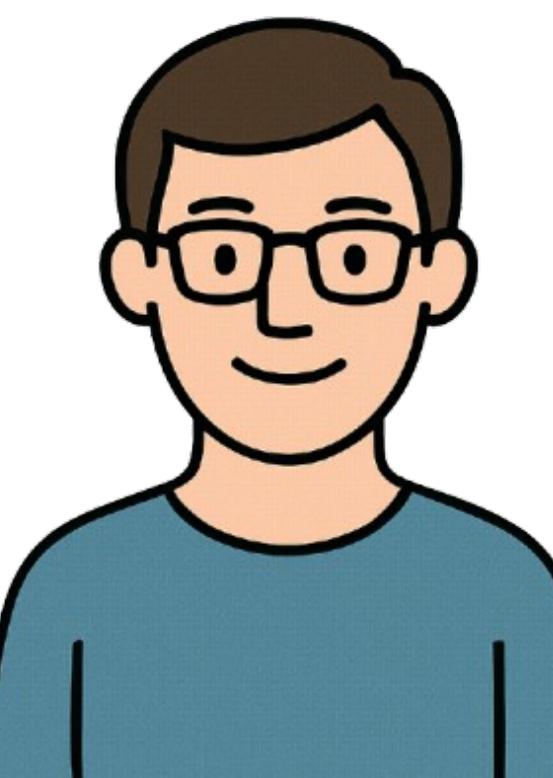
Garble(C)
 $\rightarrow \hat{C}, K$

Offline (known C)

\hat{C}

$CT(K)$

BOB



Online

Compress($K[x]$), \bar{x})

x

Size: $\text{poly}(\lambda) + |x|$

Decompress $\rightarrow L$

Eval(\hat{C}, L)
 $\rightarrow C(x)$

Existing Techniques

	Assumption	Communication	
		Offline	Online
Naive [AIKW13]		0	$ \mathbf{x} \cdot \lambda$
	RSA	$ \mathbf{x} \cdot \text{poly}(\lambda)$	$ \mathbf{x} + \log N$
	DDH/LWE	"	$ \mathbf{x} + \mathbf{x} /\lambda$

N = RSA modulus

Existing Techniques

Assumption	Communication	
	Offline	Online
Naive	0	$ \mathbf{x} \cdot \lambda$
[AIKW13]	RSA DDH/LWE	$ \mathbf{x} \cdot \text{poly}(\lambda)$ “
		$ \mathbf{x} + \log N$ $ \mathbf{x} + \mathbf{x} /\lambda$

[GS18,GOS18]: **weaker assumptions** (factoring / CDH),
but **non-black-box** cryptography

N = RSA modulus

Existing Techniques

	Assumption	Communication	
		Offline	Online
Naive [AIKW13]		0	$ \mathbf{x} \cdot \lambda$
	RSA DDH/LWE	$ \mathbf{x} \cdot \text{poly}(\lambda)$ “	$ \mathbf{x} + \log N$ $ \mathbf{x} + \mathbf{x} /\lambda$
[ABIKLV23]	RSA/iO+SSBH	0	$ \mathbf{x} + \log N$
	Bilinear DDH	0^*	$ \mathbf{x} + \mathbf{x} /\lambda$

N = RSA modulus

Existing Techniques

	Assumption	Communication Offline	Communication Online	Computation (Decompress)
Naive [AIKW13]		0	$ \mathbf{x} \cdot \lambda$	0
	RSA DDH/LWE	$ \mathbf{x} \cdot \text{poly}(\lambda)$	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations
[ABIKLV23]	RSA/iO+SSBH Bilinear DDH	0	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations

N = RSA modulus

Existing Techniques

	Assumption	Communication Offline	Communication Online	Computation (Decompress)
Naive [AIKW13]	RSA	0	$ \mathbf{x} \cdot \lambda$	0
	DDH/LWE	$ \mathbf{x} \cdot \text{poly}(\lambda)$	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations 30 min
[ABIKLV23]	RSA/iO+SSBH	0	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations
	Bilinear DDH			

(for $|\mathbf{x}| = 10^5$)

N = RSA modulus

Existing Techniques

* assuming 1Mbps

	Assumption	Communication Offline	Communication Online	Computation (Decompress)
Naive [AIKW13]	RSA	0	13 sec*	0
	DDH/LWE	$ x \cdot \text{poly}(\lambda)$	$ x + \log N$	$ x \log x $ RSA exponentiations
[ABIKLV23]	RSA/iO+SSBH	0	$ x + \log N$	30 min $ x \log x $ RSA exponentiations
	Bilinear DDH			

(for $|x| = 10^5$)

N = RSA modulus

Our work – TinyLabels

	Assumption	Communication Offline	Communication Online	Computation (Decompress)
Naive [AIKW13]	RSA	0	$ \mathbf{x} \cdot \lambda$	0
	DDH/LWE	$ \mathbf{x} \cdot \text{poly}(\lambda)$	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations
[ABIKLV23]	RSA/iO+SSBH	0	$ \mathbf{x} + \log N$	$ \mathbf{x} \log \mathbf{x} $ RSA exponentiations
	Bilinear DDH			
TinyLabels	RingLWE+RO	0^*	$ \mathbf{x} + \mathcal{R}_q $	$ \mathbf{x} \log \mathbf{x} / n$ \mathcal{R}_q multiplications

N = RSA modulus

\mathcal{R}_q = ring with degree n and modulus q

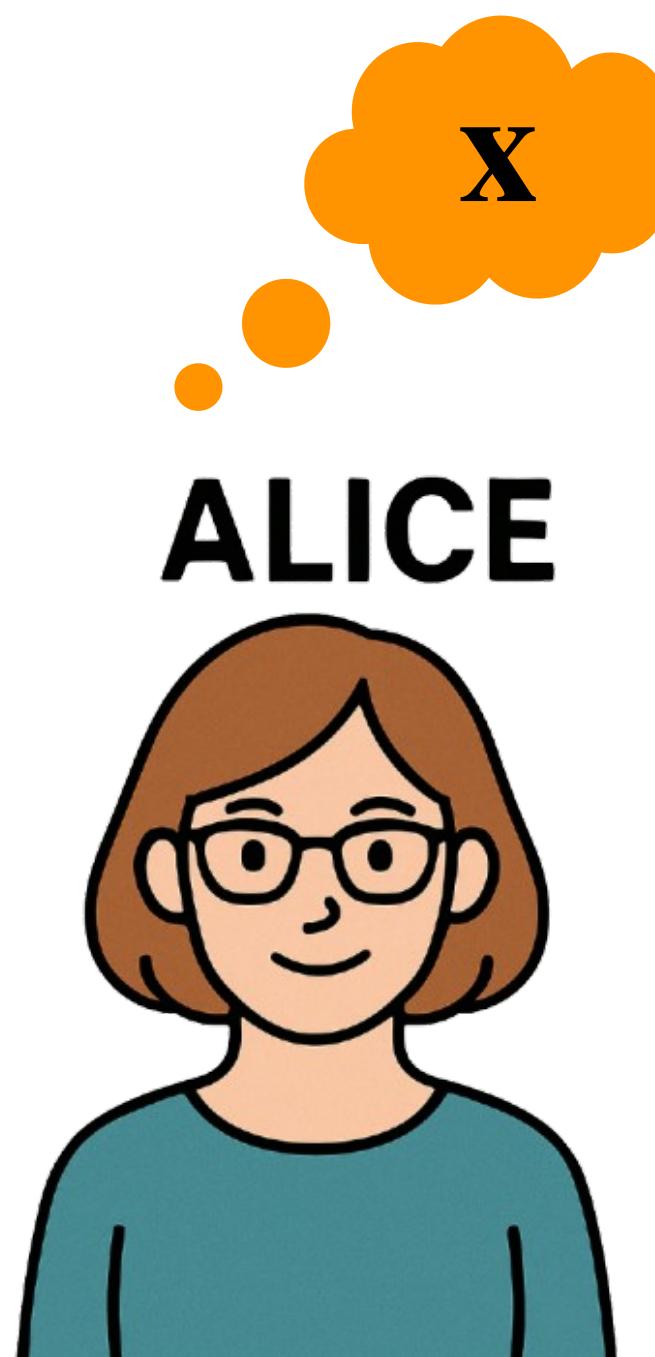
Our work – TinyLabels

	Assumption	Communication Offline	Communication Online	Computation (Decompress)
Naive [AIKW13]	RSA	0	13 sec*	0
	DDH/LWE	$ x \cdot \text{poly}(\lambda)$	$ x + \log N$	$ x \log x $ RSA exponentiations
[ABIKLV23]	RSA/iO+SSBH	0	$ x + \log N$	$ x \log x $ RSA exponentiations
	Bilinear DDH			
TinyLabels	RingLWE+RO	0*	$ x + \mathcal{R}_q $	$ x \log x / n$ \mathcal{R}_q multiplications

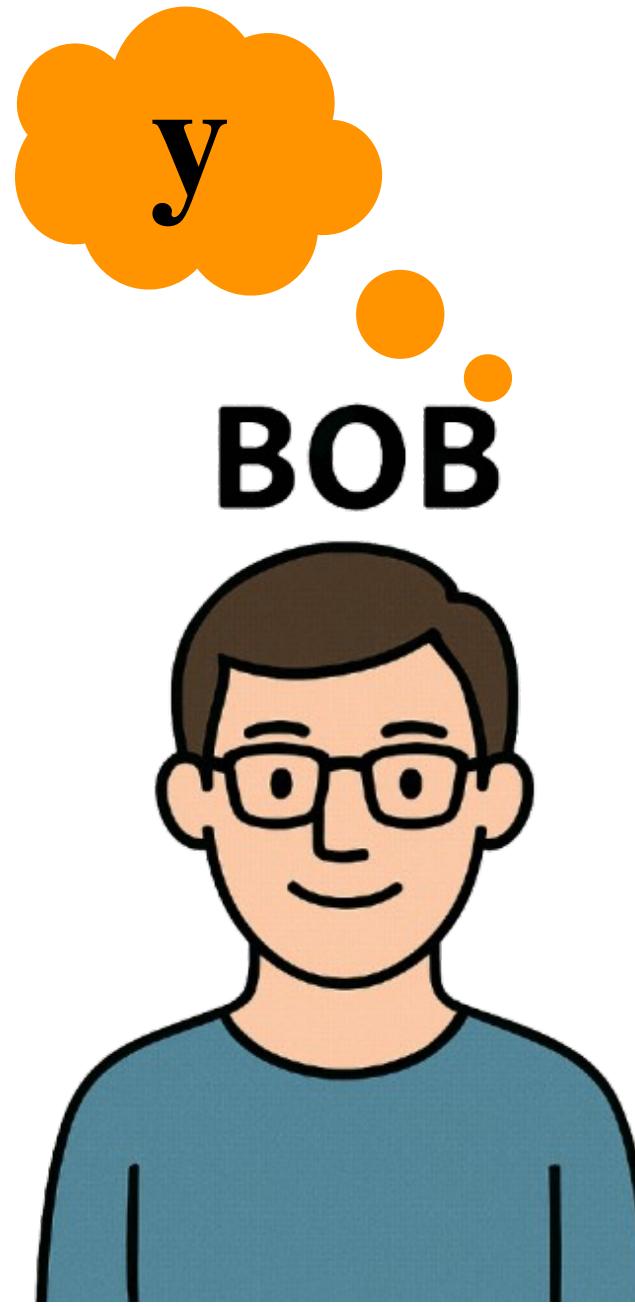
N = RSA modulus

\mathcal{R}_q = ring with degree n and modulus q

Application: Garbling with preprocessing

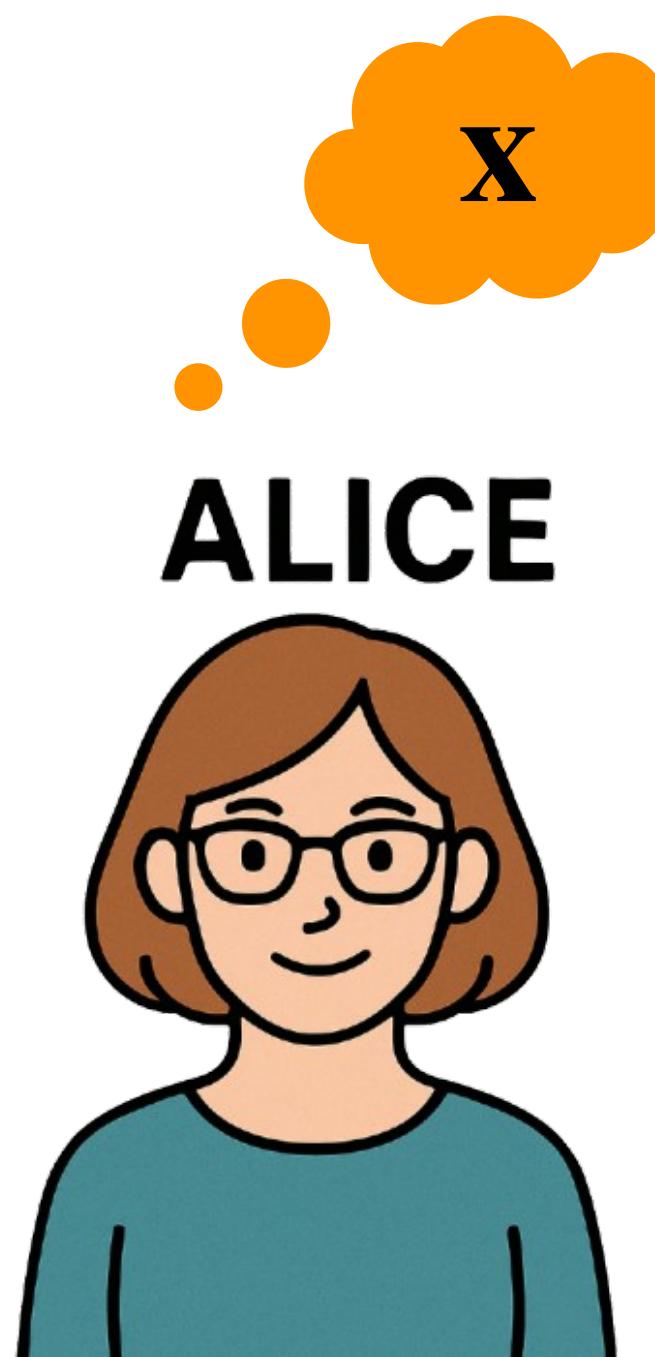


Let's analyze our
joint data securely!

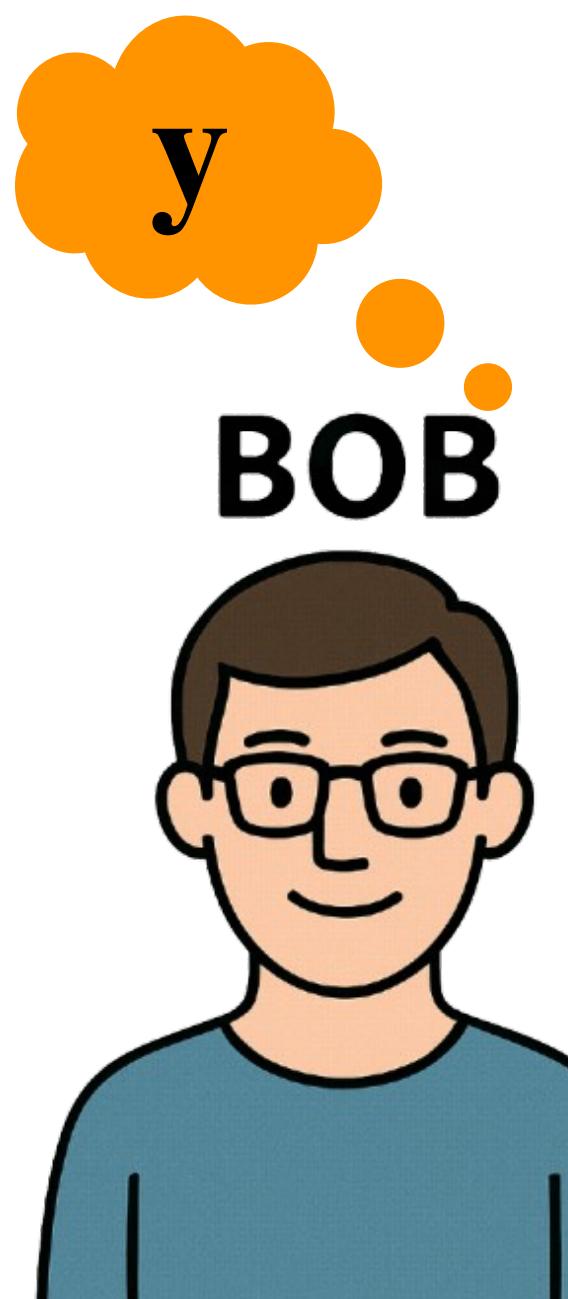


Okay, but using
which circuit C ?

Application: Garbling with preprocessing



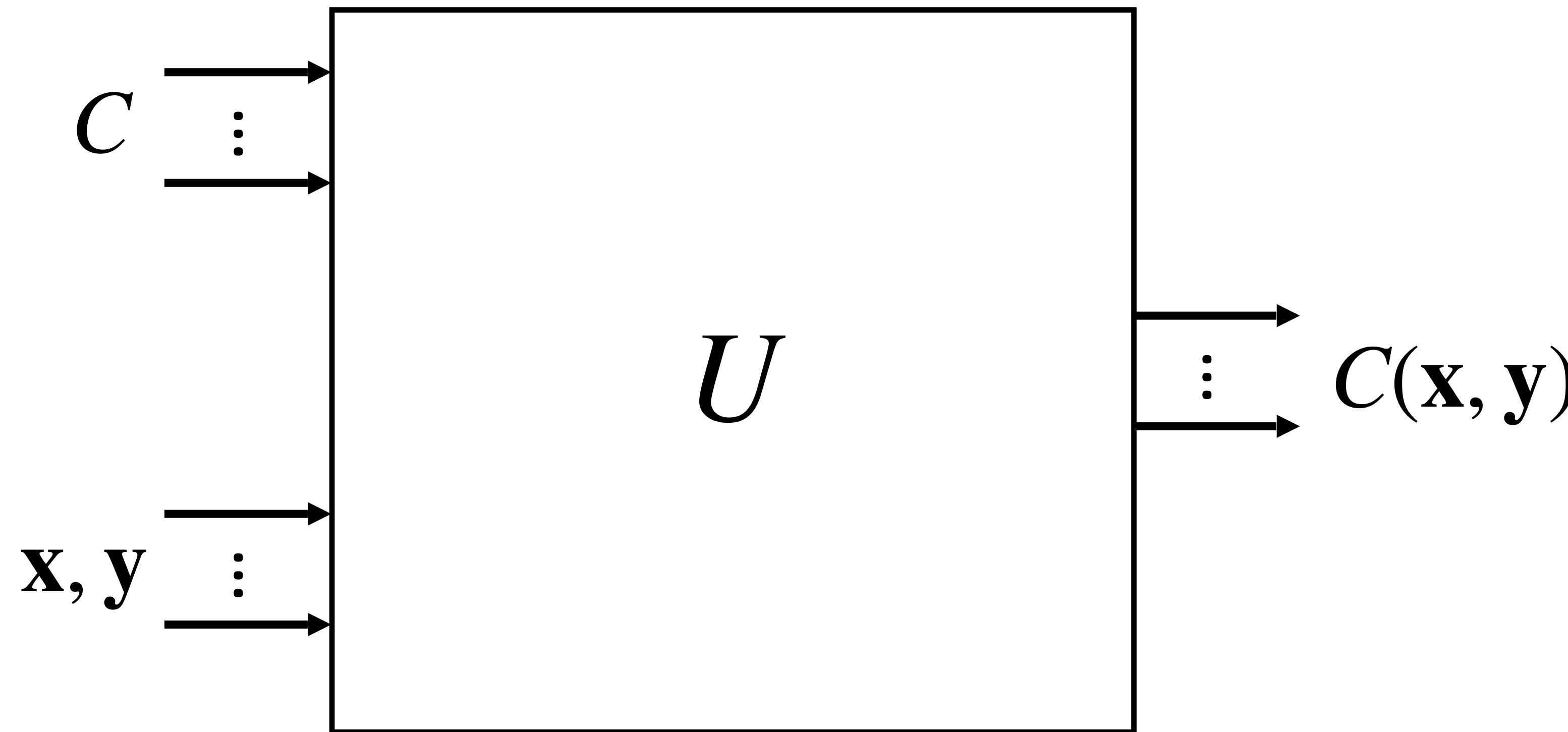
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Okay, but using which circuit C ?

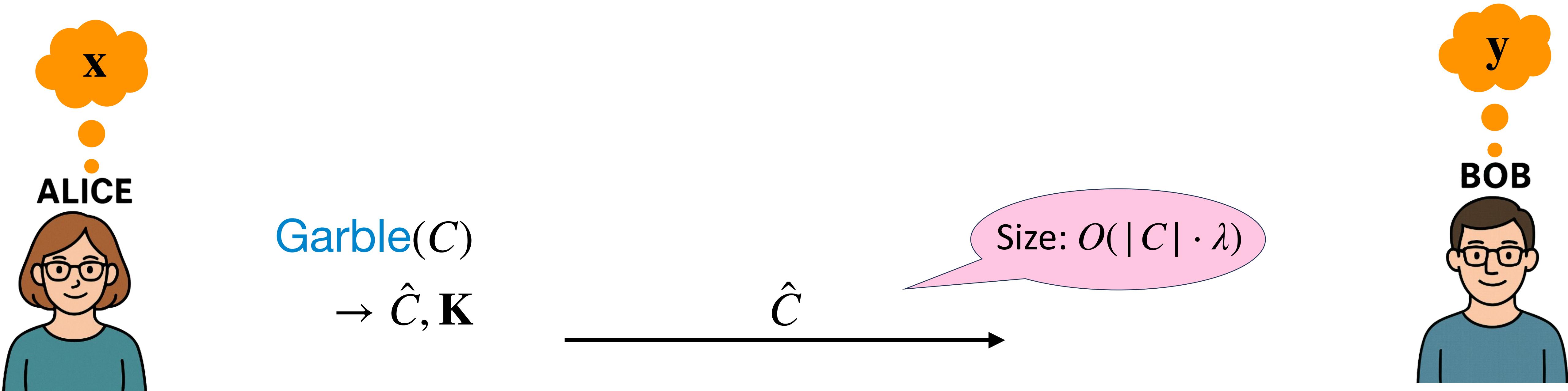
Let's discuss. In the meantime, we should preprocess with TinyLabels.

Universal Circuits

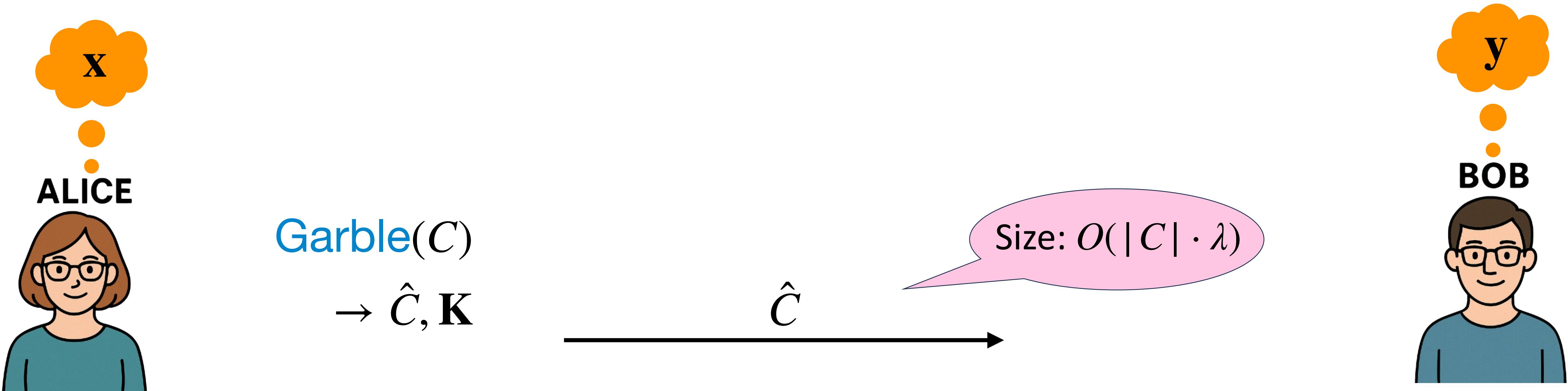


[ZYZL19]: $|U| \approx 4.5 |C| \log |C|$ AND gates

Application: Garbling with preprocessing



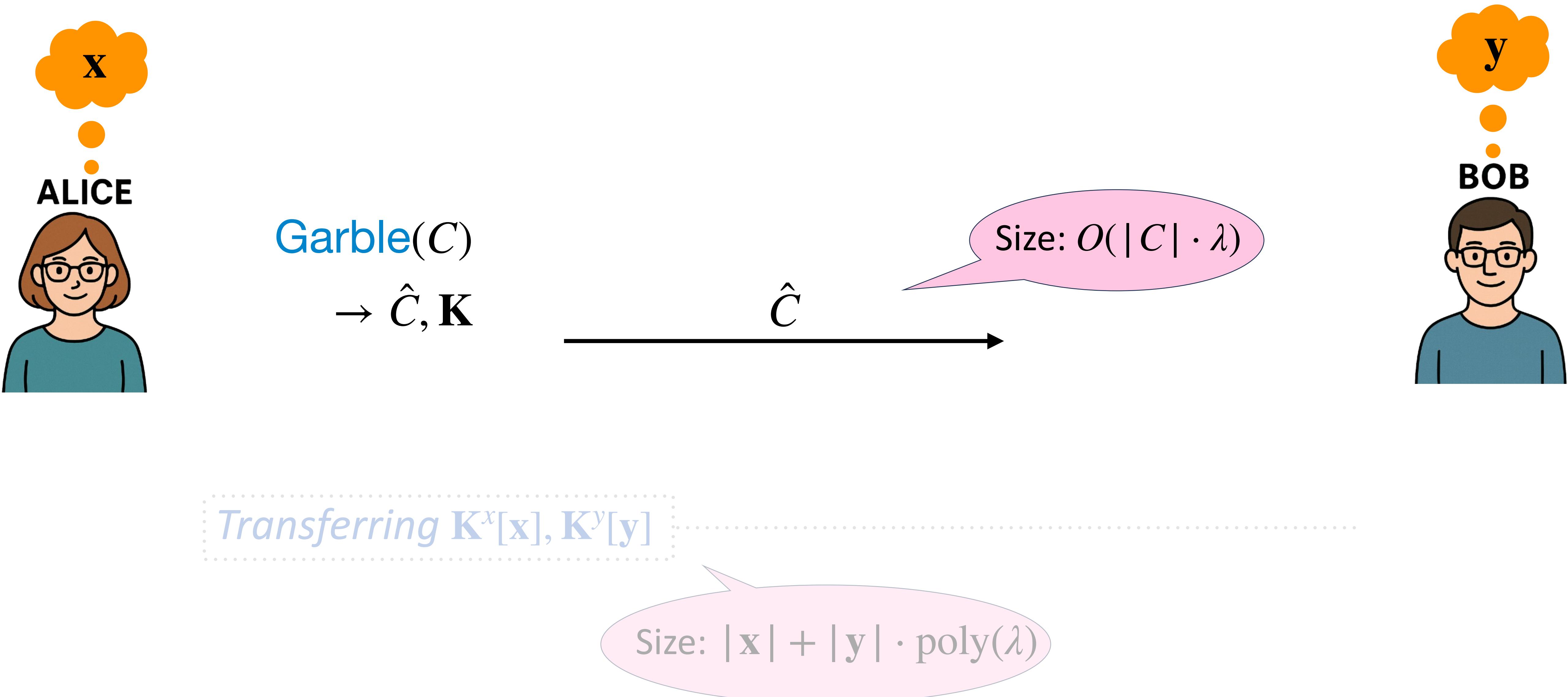
Application: Garbling with preprocessing



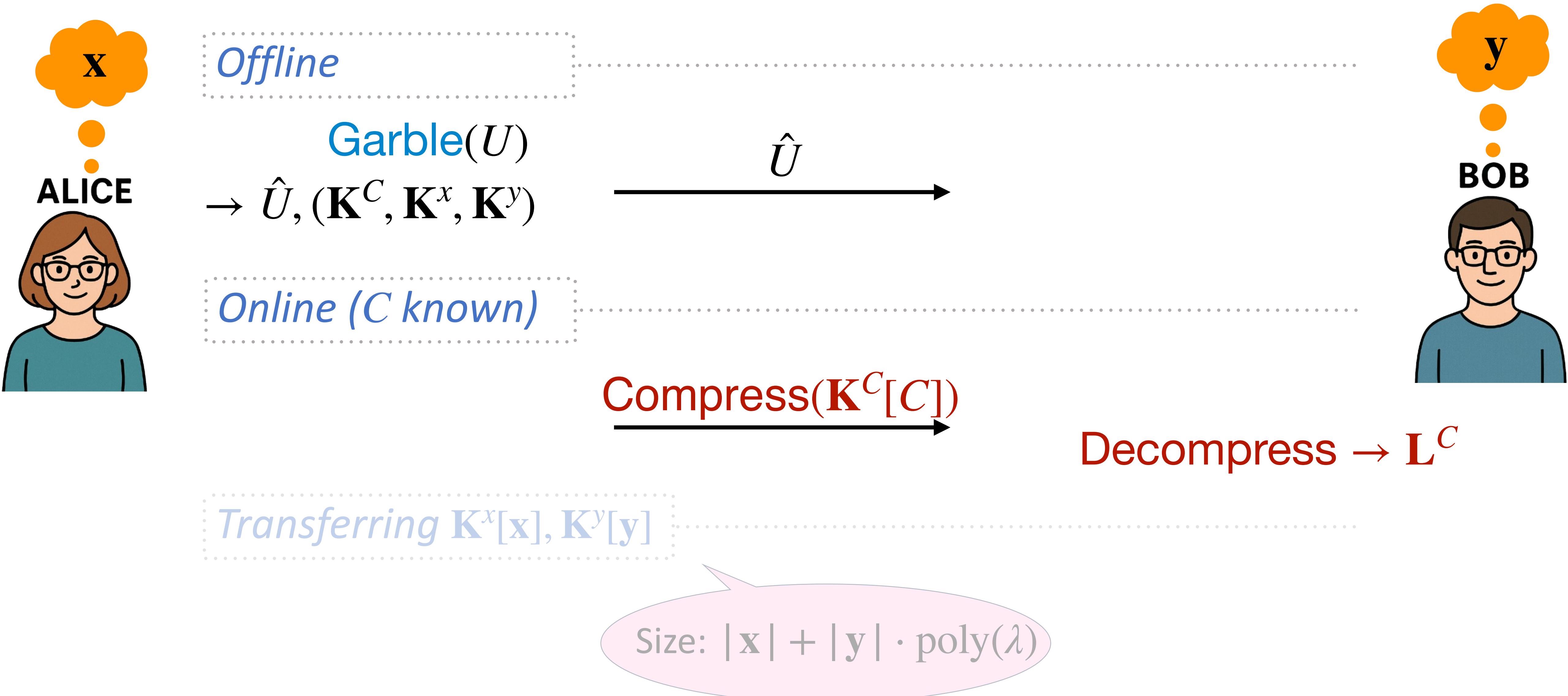
Transferring $K^x[x], K^y[y]$

Size: $|x| + |y| \cdot \text{poly}(\lambda)$

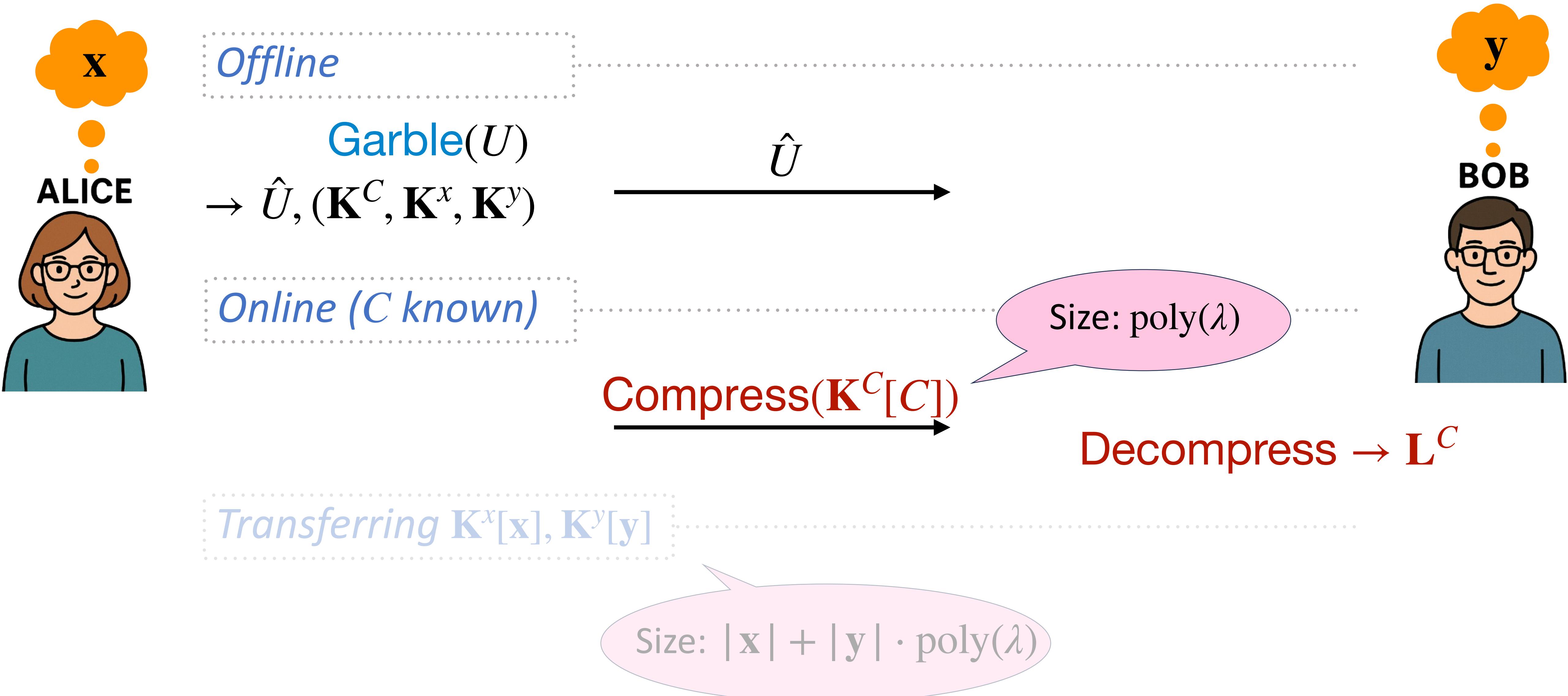
Application: Garbling with preprocessing



Application: Garbling with preprocessing



Application: Garbling with preprocessing



Main Tool: Batch-Select

Sender

Message vectors

$$\mathbf{m}_1, \mathbf{m}_2 \in \mathbb{Z}_p^w$$

$$\text{Enc}_1(\mathbf{m}_1) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{m}_2) \rightarrow \text{st}_2, \text{ct}_2$$

Main Tool: Batch-Select

Sender

Message vectors

$$\mathbf{m}_1, \mathbf{m}_2 \in \mathbb{Z}_p^w$$

$$\text{Enc}_1(\mathbf{m}_1) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{m}_2) \rightarrow \text{st}_2, \text{ct}_2$$

“Selection” vector

$$\mathbf{x} \in \mathbb{Z}_p^w$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

Main Tool: Batch-Select

	<u>Sender</u>	<u>Receiver</u>
<i>Message vectors</i>		$\text{Dec}(\text{sk}_x, \text{ct}_1, \text{ct}_2, x)$
$\mathbf{m}_1, \mathbf{m}_2 \in \mathbb{Z}_p^w$	$\text{Enc}_1(\mathbf{m}_1) \rightarrow \text{st}_1, \text{ct}_1$ $\text{Enc}_2(\mathbf{m}_2) \rightarrow \text{st}_2, \text{ct}_2$	$\rightarrow \mathbf{m}_1 \odot x + \mathbf{m}_2$
<i>“Selection” vector</i>	$\text{KeyGen}(\text{st}_1, \text{st}_2, x) \rightarrow \text{sk}_x$	$= \begin{pmatrix} \mathbf{m}_1[1] \cdot x_1 + \mathbf{m}_2[1] \\ \vdots \\ \mathbf{m}_w[w] \cdot x_w + \mathbf{m}_2[w] \end{pmatrix}$

Main Tool: Batch-Select

	<u>Sender</u>	<u>Receiver</u>
<i>Message vectors</i> $\mathbf{m}_1, \mathbf{m}_2 \in \mathbb{Z}_p^w$	$\text{Enc}_1(\mathbf{m}_1) \rightarrow \text{st}_1, \text{ct}_1$ $\text{Enc}_2(\mathbf{m}_2) \rightarrow \text{st}_2, \text{ct}_2$	$\text{Dec}(\text{sk}_{\mathbf{x}}, \text{ct}_1, \text{ct}_2, \mathbf{x})$ $\rightarrow \mathbf{m}_1 \odot \mathbf{x} + \mathbf{m}_2$
<i>“Selection” vector</i> $\mathbf{x} \in \mathbb{Z}_p^w$	$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$	$= \begin{pmatrix} \mathbf{m}_1[1] \cdot x_1 + \mathbf{m}_2[1] \\ \vdots \\ \mathbf{m}_w[w] \cdot x_w + \mathbf{m}_2[w] \end{pmatrix}$

Succinctness: $|\text{sk}_{\mathbf{x}}| = \text{poly}(\lambda)$

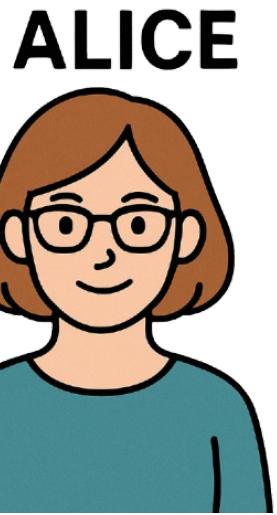
Main Tool: Batch-Select

	<u>Sender</u>	<u>Receiver</u>
<i>Message vectors</i> $\mathbf{m}_1, \mathbf{m}_2 \in \mathbb{Z}_p^w$	$\text{Enc}_1(\mathbf{m}_1) \rightarrow \text{st}_1, \text{ct}_1$ $\text{Enc}_2(\mathbf{m}_2) \rightarrow \text{st}_2, \text{ct}_2$	$\text{Dec}(\text{sk}_{\mathbf{x}}, \text{ct}_1, \text{ct}_2, \mathbf{x})$ $\rightarrow \mathbf{m}_1 \odot \mathbf{x} + \mathbf{m}_2$
<i>“Selection” vector</i> $\mathbf{x} \in \mathbb{Z}_p^w$	$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$	$= \begin{pmatrix} \mathbf{m}_1[1] \cdot x_1 + \mathbf{m}_2[1] \\ \vdots \\ \mathbf{m}_w[w] \cdot x_w + \mathbf{m}_2[w] \end{pmatrix}$

**Simulation
Security:**

For all $\mathbf{m}_1, \mathbf{m}_2, \mathbf{x}$:
 $(\text{sk}_{\mathbf{x}}, \text{ct}_1, \text{ct}_2) \approx \text{Sim}(\mathbf{m}_1 \odot \mathbf{x} + \mathbf{m}_2, \mathbf{x})$

Compressing Labels using Batch-Select



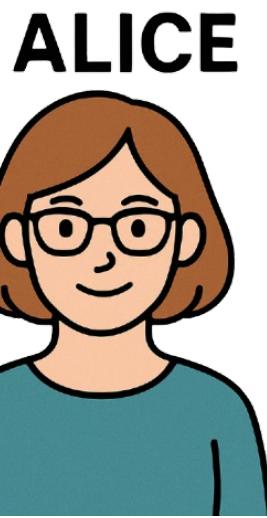
GC input keys

$$\mathbf{K}[0], \mathbf{K}[1] \in \mathbb{Z}_p^w$$

$$\text{Enc}_1(\mathbf{K}[1] - \mathbf{K}[0]) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{K}[0]) \rightarrow \text{st}_2, \text{ct}_2$$

Compressing Labels using Batch-Select



GC input keys

$$\mathbf{K}[0], \mathbf{K}[1] \in \mathbb{Z}_p^w$$

$$\text{Enc}_1(\mathbf{K}[1] - \mathbf{K}[0]) \rightarrow \text{st}_1, \text{ct}_1$$

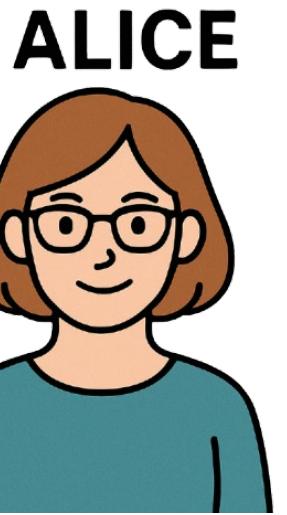
$$\text{Enc}_2(\mathbf{K}[0]) \rightarrow \text{st}_2, \text{ct}_2$$

Input

$$\mathbf{x} \in \{0,1\}^w$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

Compressing Labels using Batch-Select



GC input keys

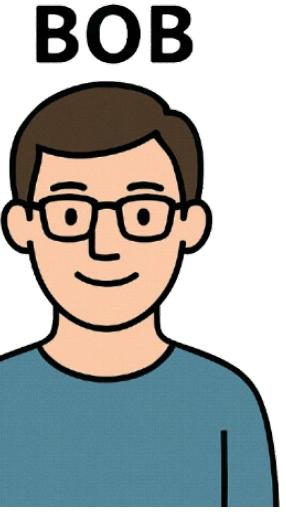
$$\mathbf{K}[0], \mathbf{K}[1] \in \mathbb{Z}_p^w$$

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$$\text{Enc}_1(\mathbf{K}[1] - \mathbf{K}[0]) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{K}[0]) \rightarrow \text{st}_2, \text{ct}_2$$



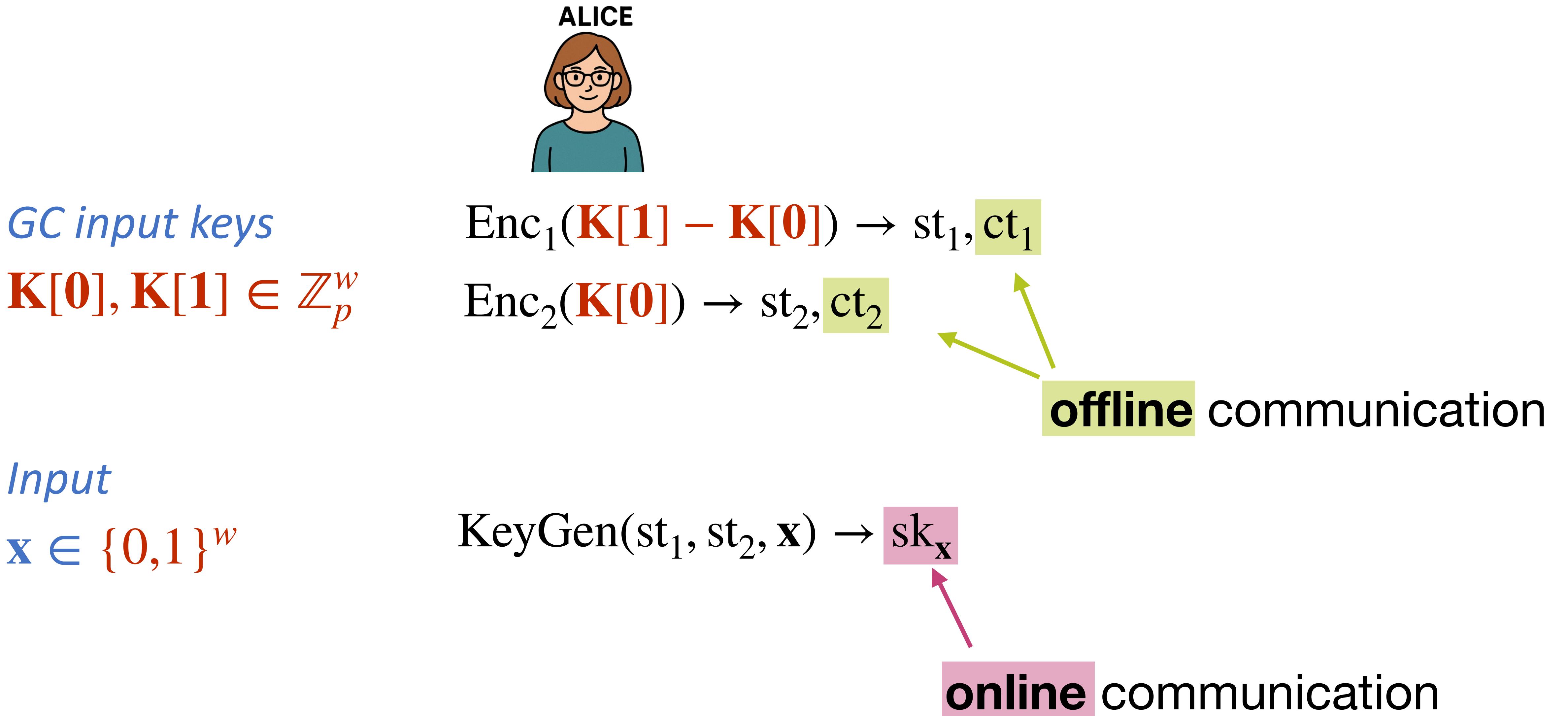
$$\text{Dec}(\text{sk}_{\mathbf{x}}, \text{ct}_1, \text{ct}_2, \mathbf{x})$$

$$\rightarrow (\mathbf{K}[1] - \mathbf{K}[0]) \odot \mathbf{x} + \mathbf{K}[0]$$

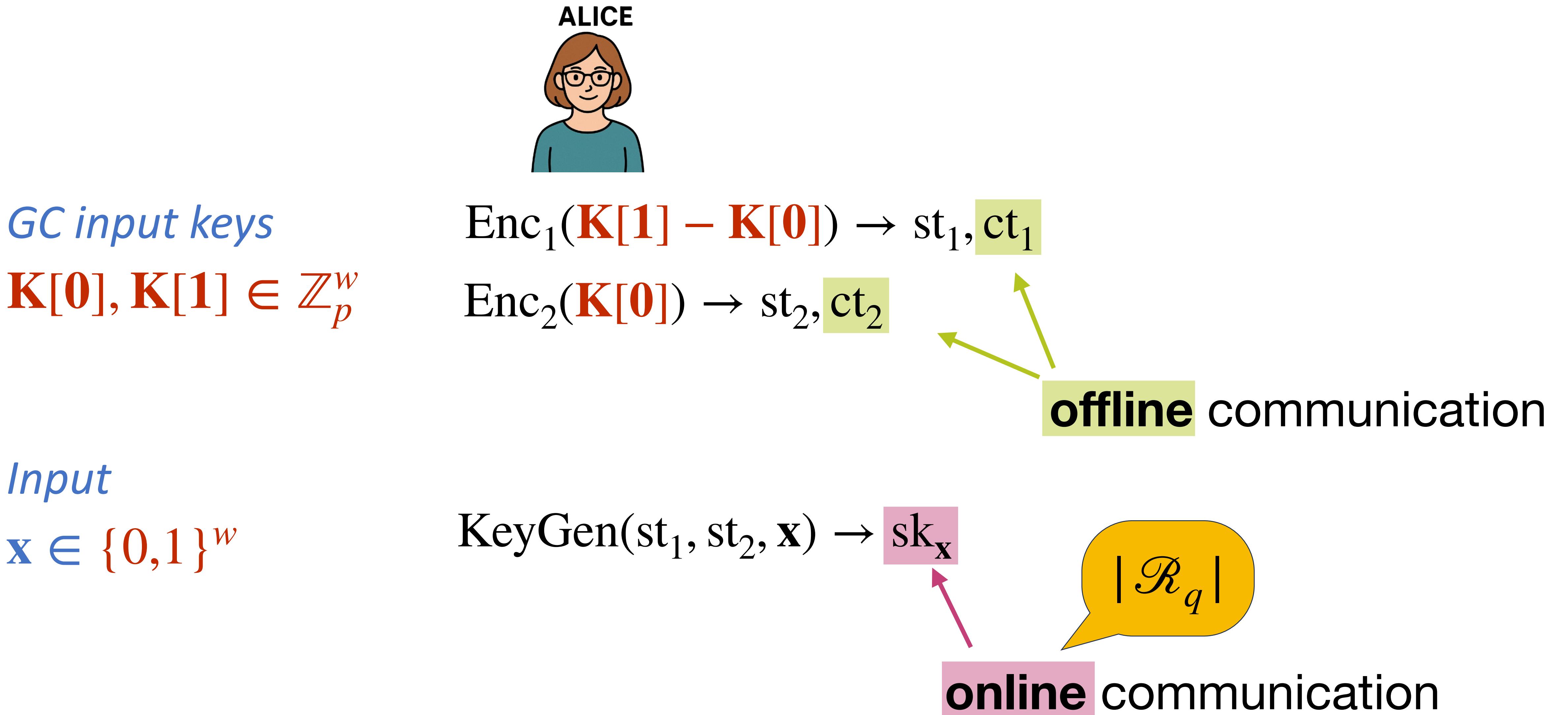
$$= \begin{pmatrix} \mathbf{K}[x_1] \\ \vdots \\ \mathbf{K}[x_w] \end{pmatrix} = \mathbf{K}[\mathbf{x}]$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

Compressing Labels using Batch-Select

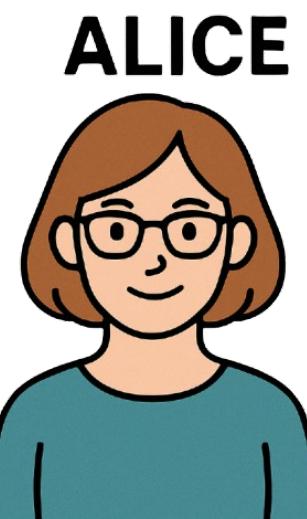


Compressing Labels using Batch-Select



(Assuming our construction from RingLWE)

Compressing Labels using Batch-Select



GC input keys

$$\mathbf{K}[0], \mathbf{K}[1] \in \mathbb{Z}_p^w$$

Input

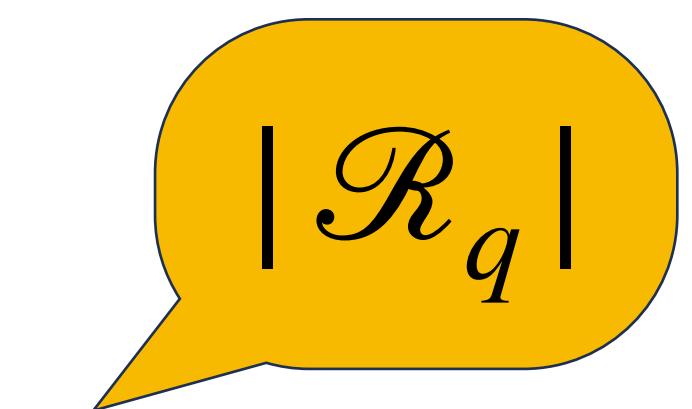
$$\mathbf{x} \in \{0,1\}^w$$

$$\text{Enc}_1(\mathbf{K}[1] - \mathbf{K}[0]) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{K}[0]) \rightarrow \text{st}_2, \text{ct}_2$$

offline communication

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

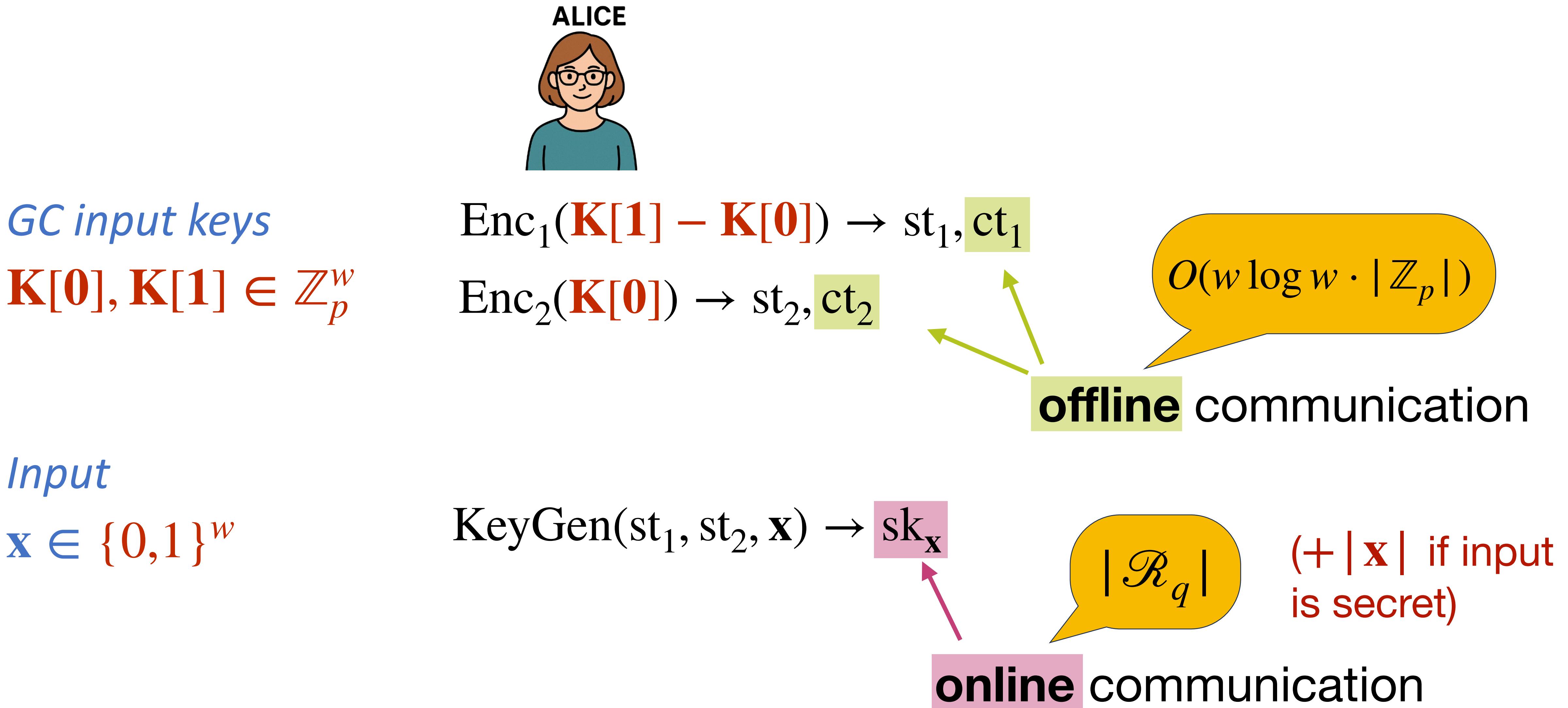


(+ | \mathbf{x} | if input
is secret)

online communication

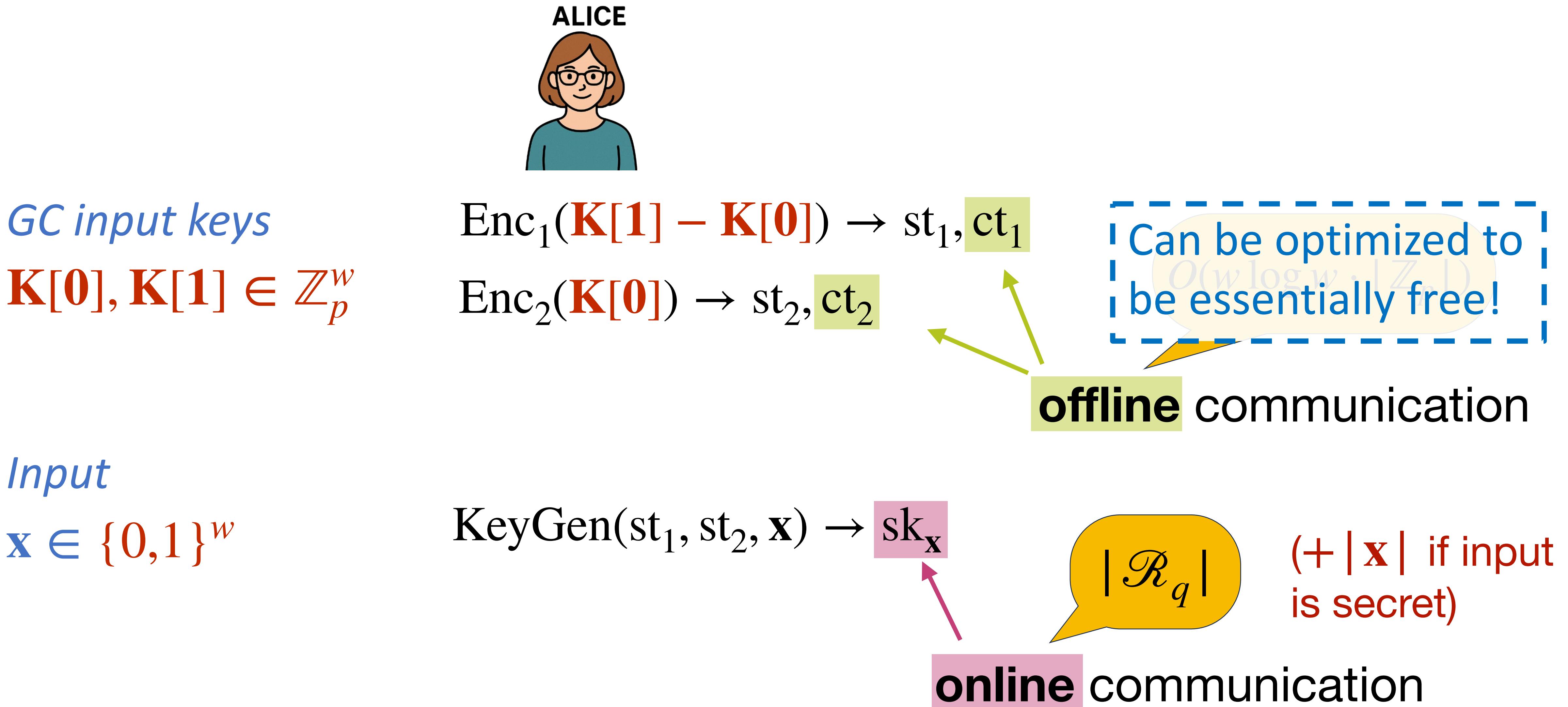
(Assuming our construction from RingLWE)

Compressing Labels using Batch-Select



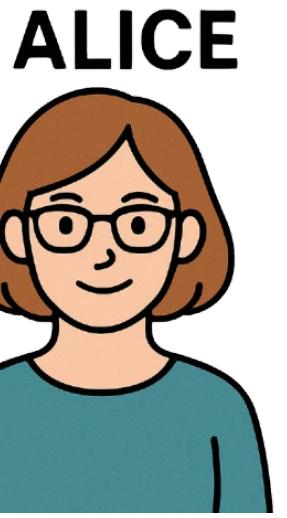
(Assuming our construction from RingLWE)

Compressing Labels using Batch-Select



(Assuming our construction from RingLWE)

Optimization 1: Free-XOR keys



GC input keys

$$\mathbf{K[0]}$$

$$\mathbf{K[1]} = \mathbf{K[0]} + \Delta \cdot \mathbf{1}_w$$

$$\text{Enc}_1(\mathbf{K[1]} - \mathbf{K[0]}) \rightarrow \text{st}_1, \text{ct}_1$$

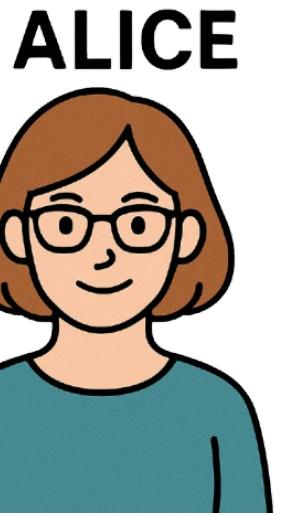
$$\text{Enc}_2(\mathbf{K[0]}) \rightarrow \text{st}_2, \text{ct}_2$$

Input

$$\mathbf{x} \in \{0,1\}^w$$

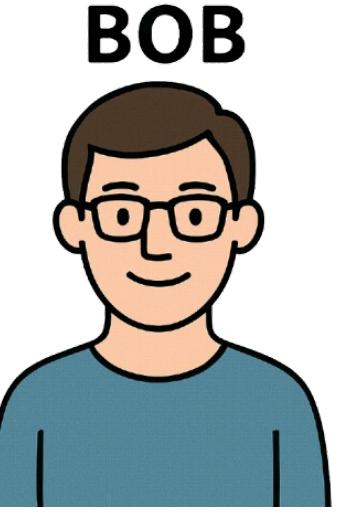
$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

Optimization 1: Free-XOR keys



GC input keys

$$\begin{aligned} \mathbf{K[0]} \\ \mathbf{K[1]} = \mathbf{K[0]} + \Delta \cdot \mathbf{1}_w \end{aligned}$$



$$\text{Enc}_1(\Delta \cdot \mathbf{1}_w) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{Enc}_2(\mathbf{K[0]}) \rightarrow \text{st}_2, \text{ct}_2$$

$$\begin{aligned} \text{Dec}(\text{sk}_x, \text{ct}_1, \text{ct}_2, x) \\ \rightarrow (\Delta \cdot \mathbf{1}_w) \odot x + \mathbf{K[0]} \\ = \mathbf{K[x]} \end{aligned}$$

Input

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

$$\text{KeyGen(st}_1, \text{st}_2, x) \rightarrow \text{sk}_x$$

Optimization 1: Free-XOR keys

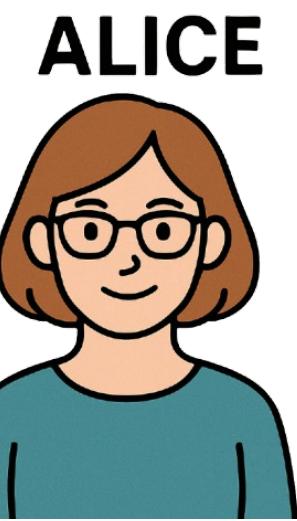
GC input keys

$$\mathbf{K[0]}$$

$$\mathbf{K[1]} = \mathbf{K[0]} + \Delta \cdot \mathbf{1}_w$$

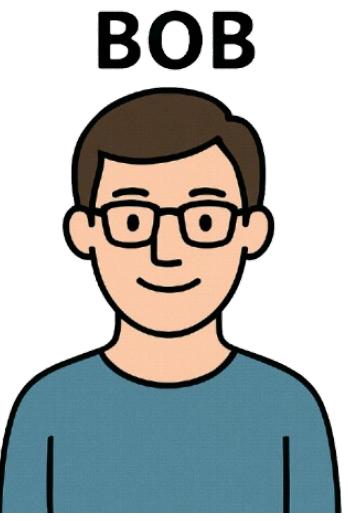
Input

$$\mathbf{x} \in \{0,1\}^w$$



$$\text{Enc}_1(\Delta \cdot \mathbf{1}_w) \rightarrow \text{st}_1, \text{ct}_1$$

Reusable!

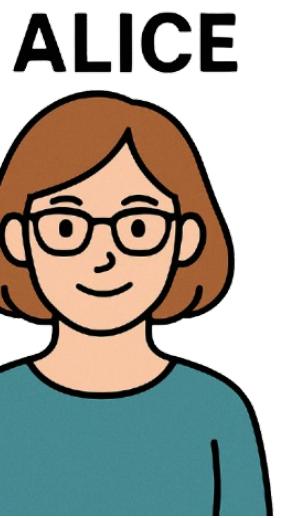


$$\text{Dec}(\text{sk}_x, \text{ct}_1, \text{ct}_2, \mathbf{x})$$

$$\begin{aligned} \text{Enc}_2(\mathbf{K[0]}) \rightarrow \text{st}_2, \text{ct}_2 \\ \rightarrow (\Delta \cdot \mathbf{1}_w) \odot \mathbf{x} + \mathbf{K[0]} \\ = \mathbf{K[x]} \end{aligned}$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_x$$

Optimization 2: Free encryption of random keys



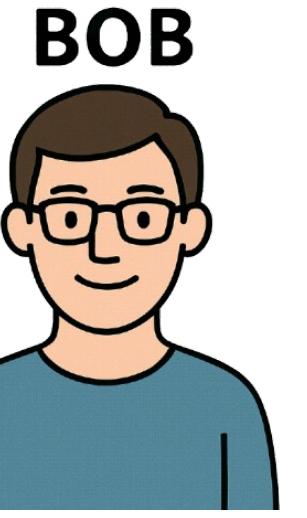
GC input keys

$$\mathbf{K[0]} \leftarrow \$$$

$$\mathbf{K[1]} = \mathbf{K[0]} + \Delta \cdot \mathbf{1}_w$$

Input

$$\mathbf{x} \in \{0,1\}^w$$



$$\text{Enc}_1(\Delta \cdot \mathbf{1}_w) \rightarrow \text{st}_1, \text{ct}_1$$

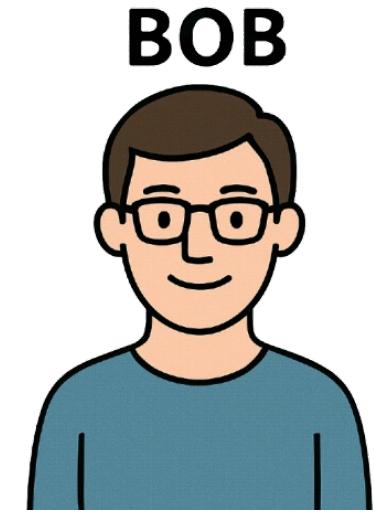
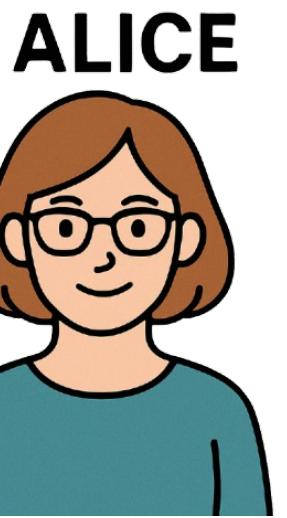
$$\text{Enc}_2(\mathbf{K[0]}) \rightarrow \text{st}_2, \text{ct}_2$$

$$\text{Dec}(\text{sk}_{\mathbf{x}}, \text{ct}_1, \text{ct}_2, \mathbf{x})$$

$$\begin{aligned} & \rightarrow (\Delta \cdot \mathbf{1}_w) \odot \mathbf{x} + \mathbf{K[0]} \\ & = \mathbf{K[x]} \end{aligned}$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, \mathbf{x}) \rightarrow \text{sk}_{\mathbf{x}}$$

Optimization 2: Free encryption of random keys



GC input keys

$$\mathbf{K[0]} \leftarrow \$$$

$$\mathbf{K[1]} = \mathbf{K[0]} + \Delta \cdot \mathbf{1}_w$$

$$\text{Enc}_1(\Delta \cdot \mathbf{1}_w) \rightarrow \text{st}_1, \text{ct}_1$$

$$\text{ct}_2 \leftarrow \text{RO}(\text{seed})$$

$$\mathbf{K[0]}, \text{st}_2 \leftarrow \text{Dec}(\text{ct}_2)$$

$$\text{Dec}(\text{sk}_x, \text{ct}_1, \text{ct}_2, x)$$

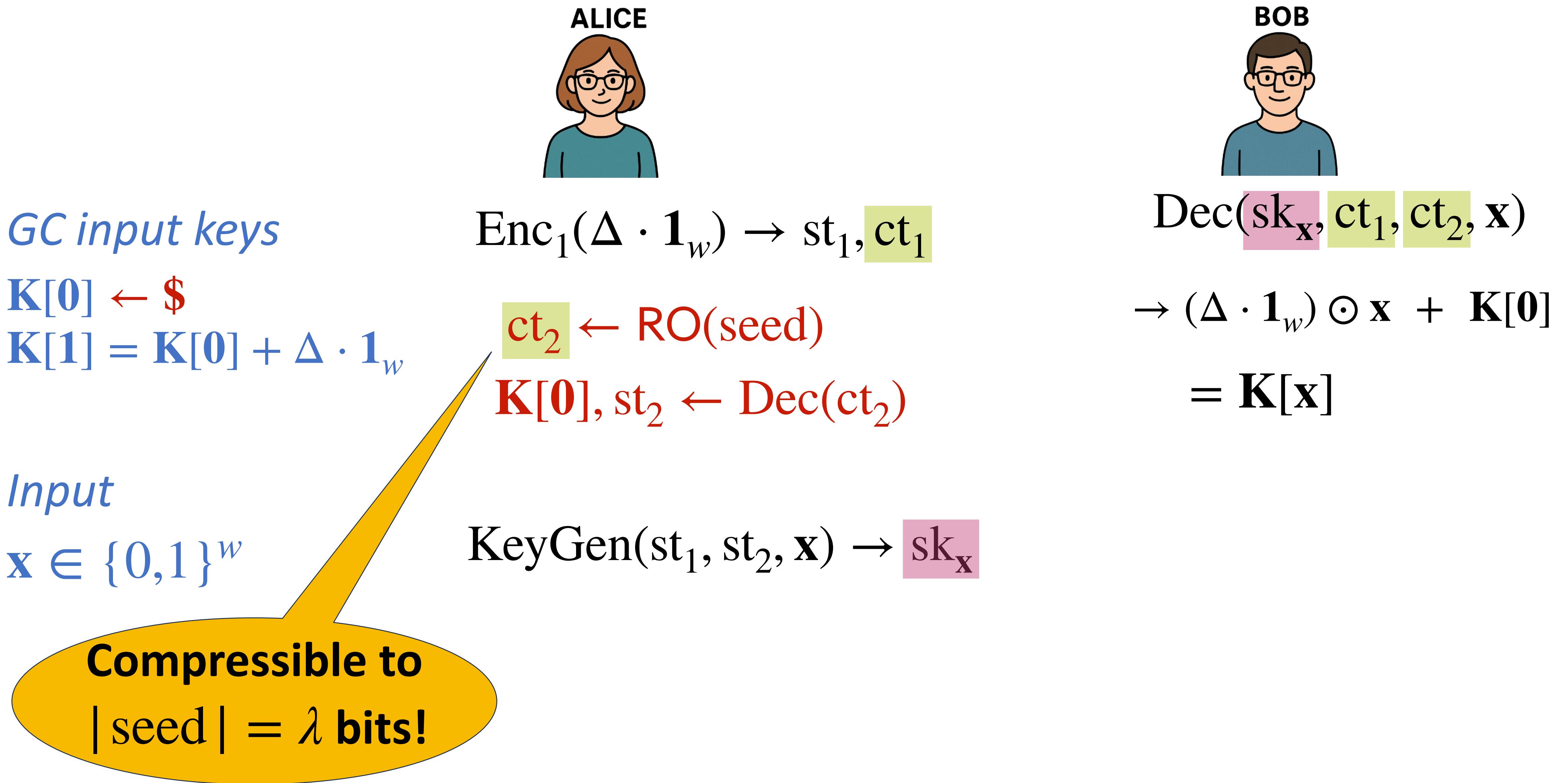
$$\begin{aligned} & \rightarrow (\Delta \cdot \mathbf{1}_w) \odot x + \mathbf{K[0]} \\ & = \mathbf{K[x]} \end{aligned}$$

Input

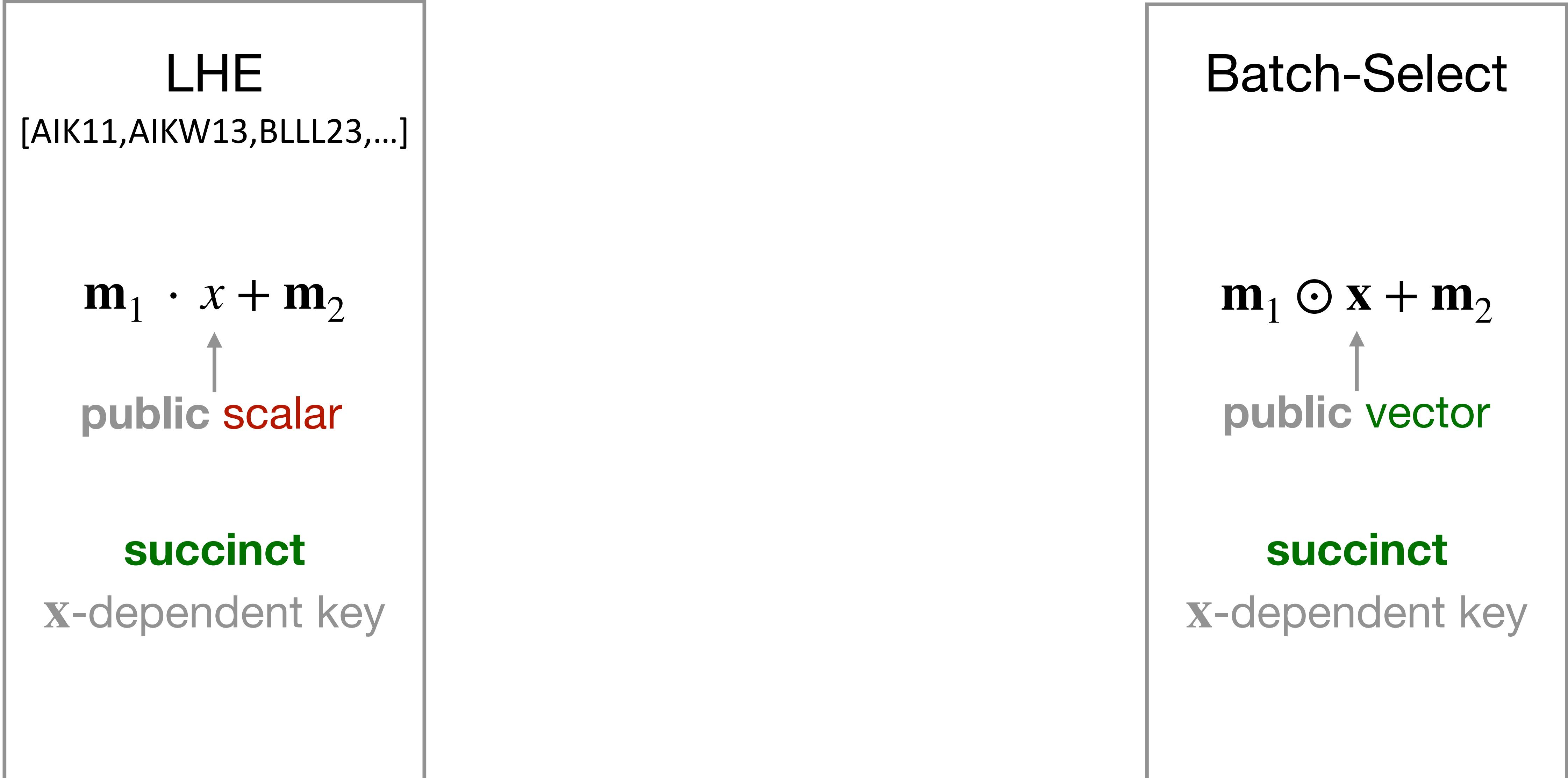
$$x \in \{0,1\}^w$$

$$\text{KeyGen}(\text{st}_1, \text{st}_2, x) \rightarrow \text{sk}_x$$

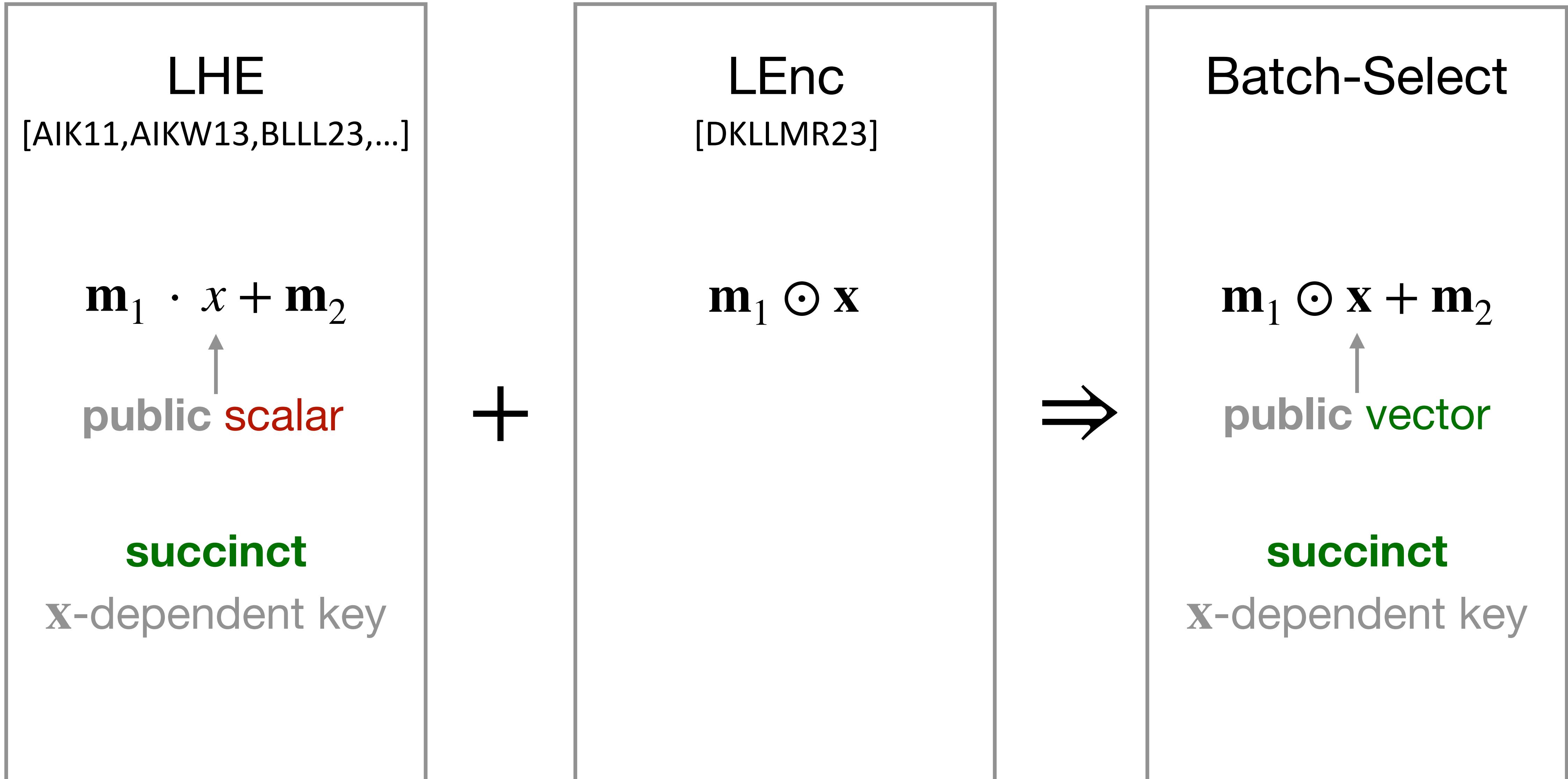
Optimization 2: Free encryption of random keys



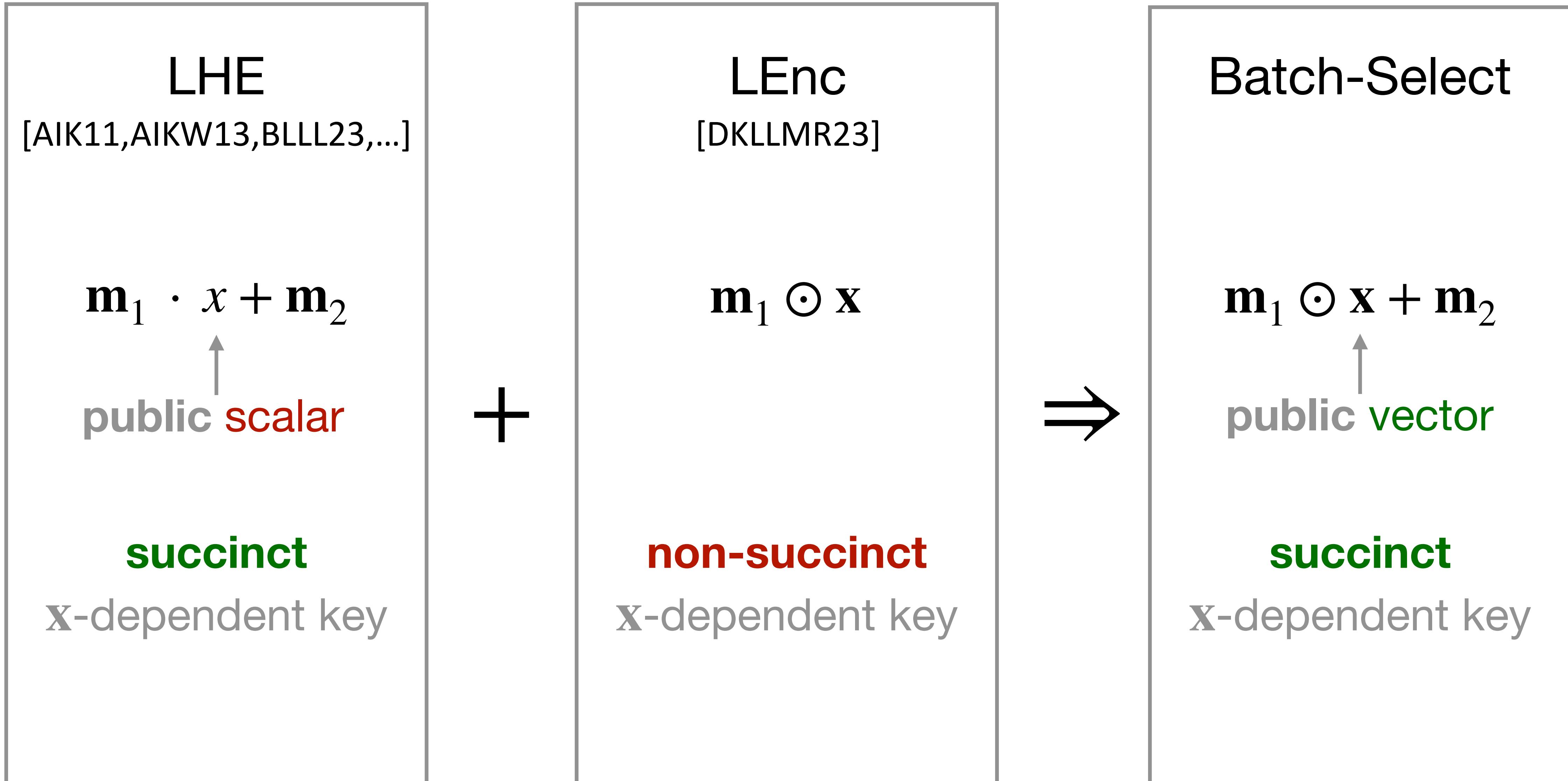
Constructing Batch-Select from RingLWE



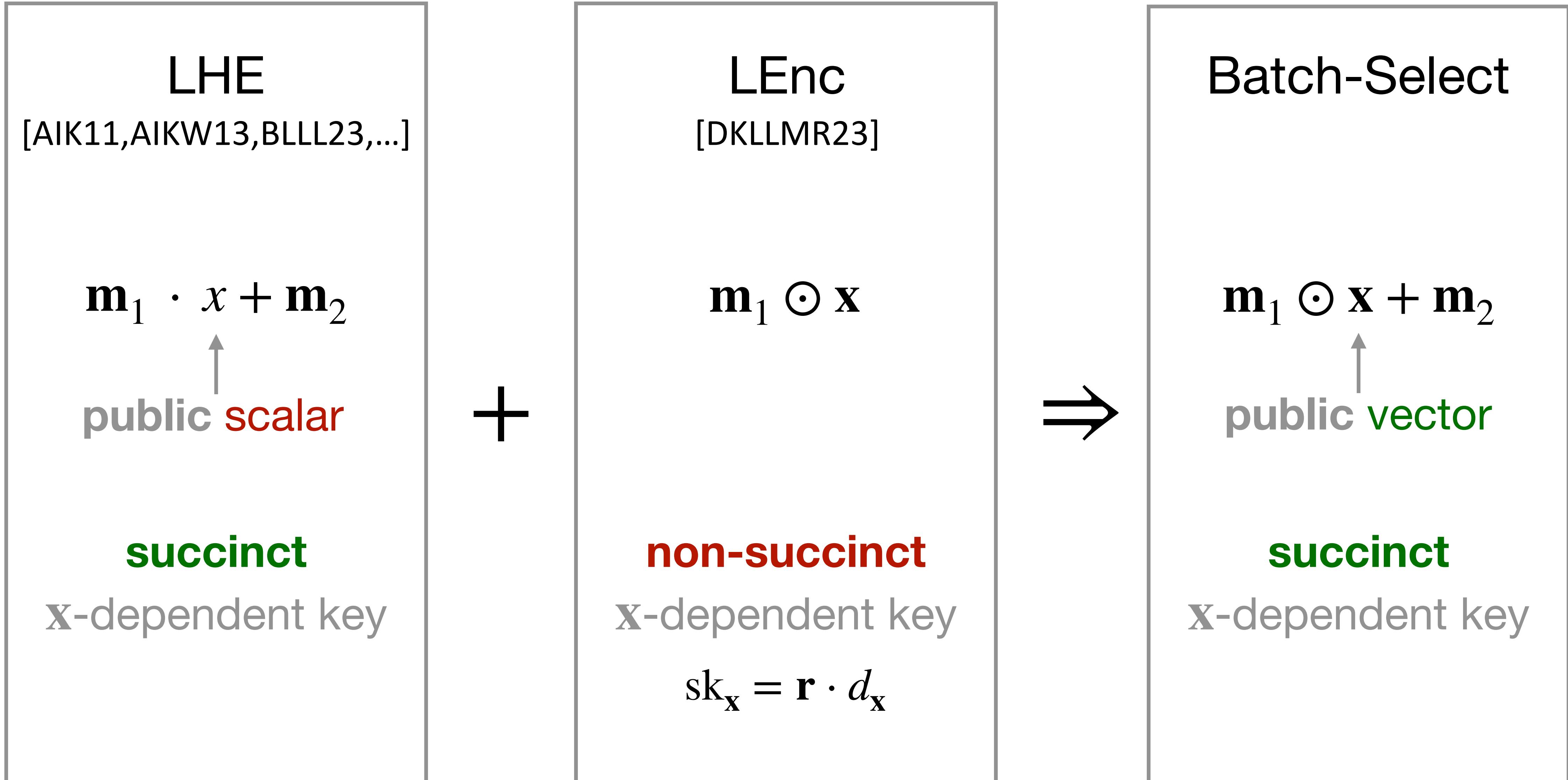
Constructing Batch-Select from RingLWE



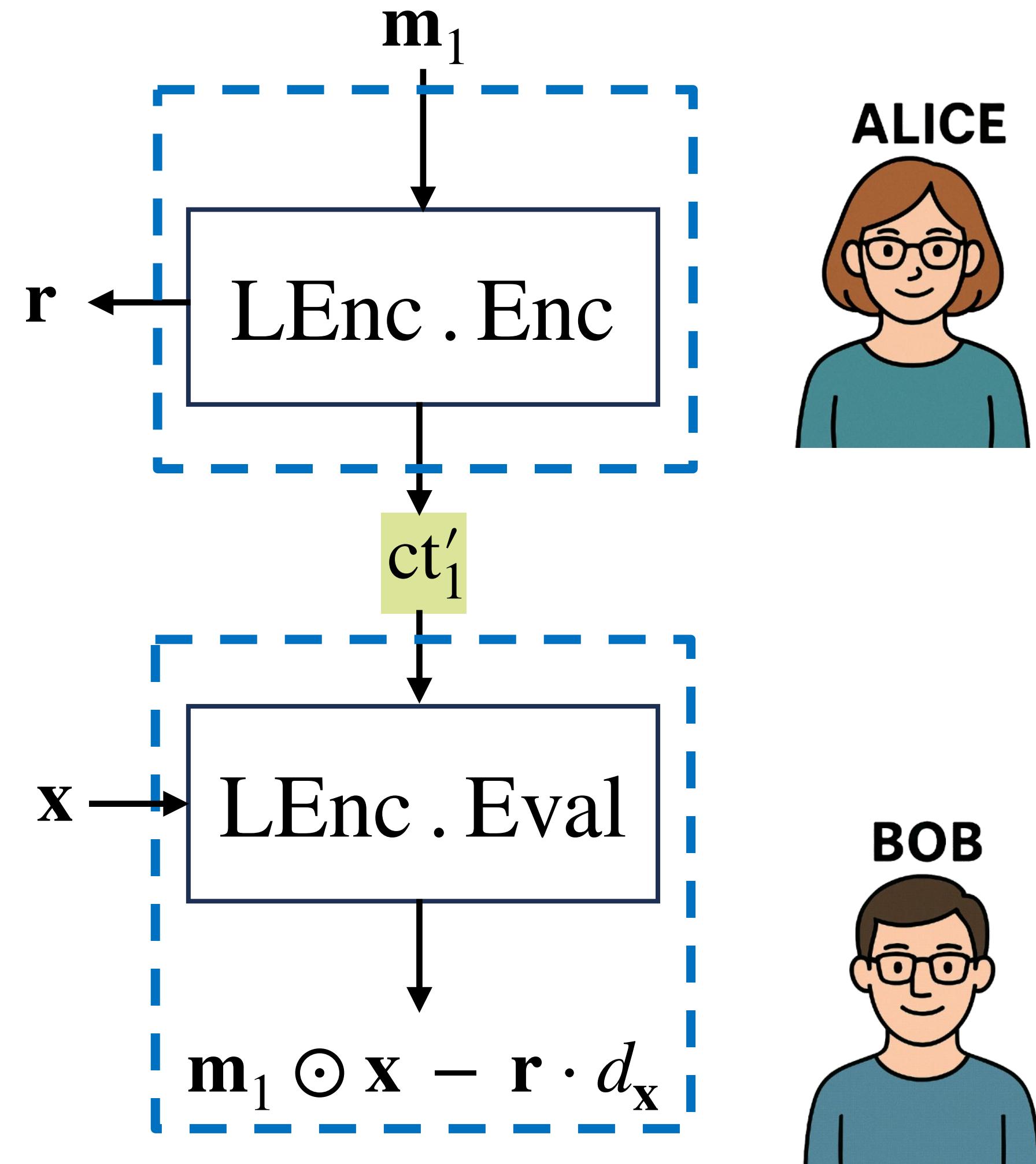
Constructing Batch-Select from RingLWE



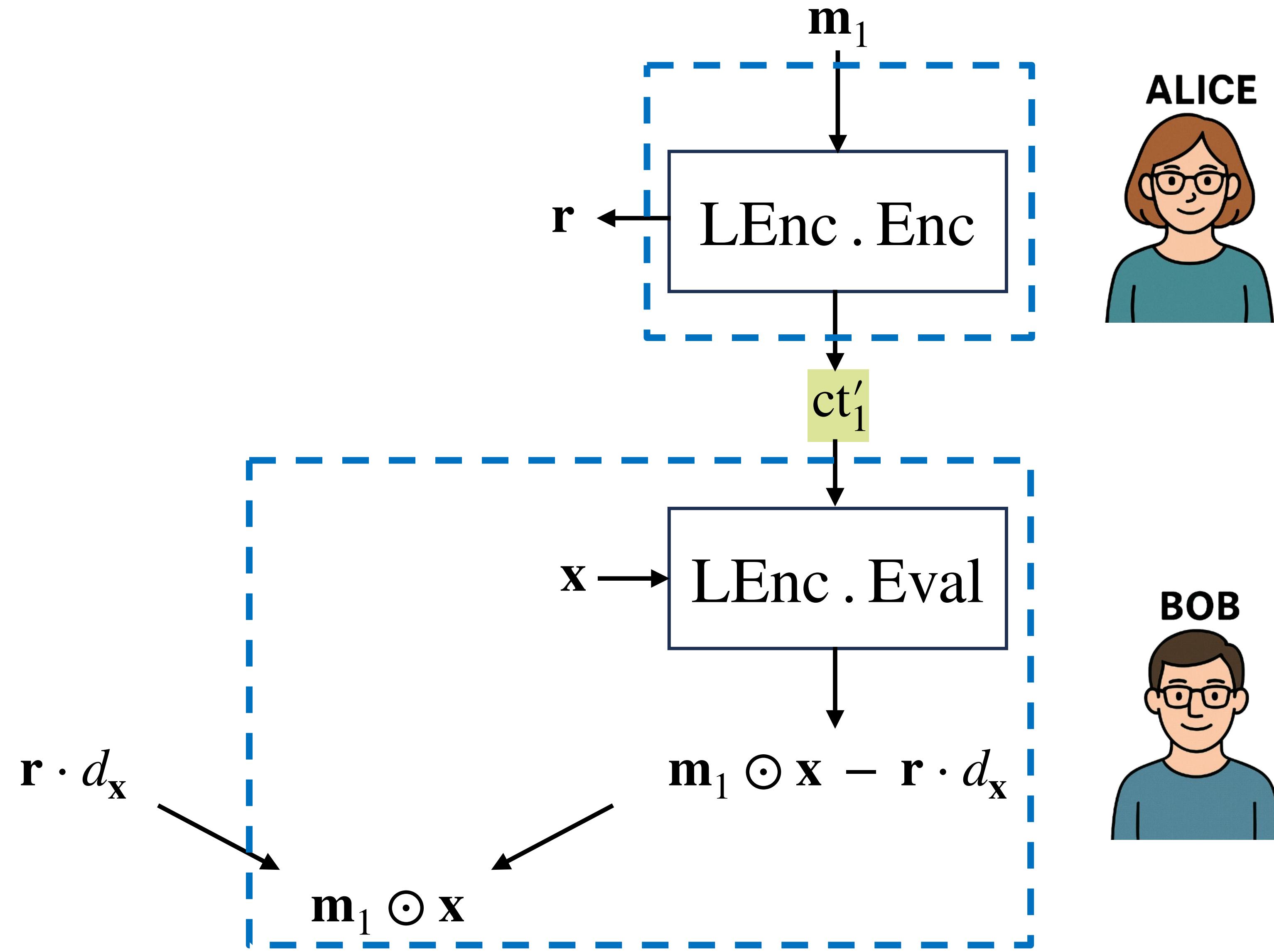
Constructing Batch-Select from RingLWE



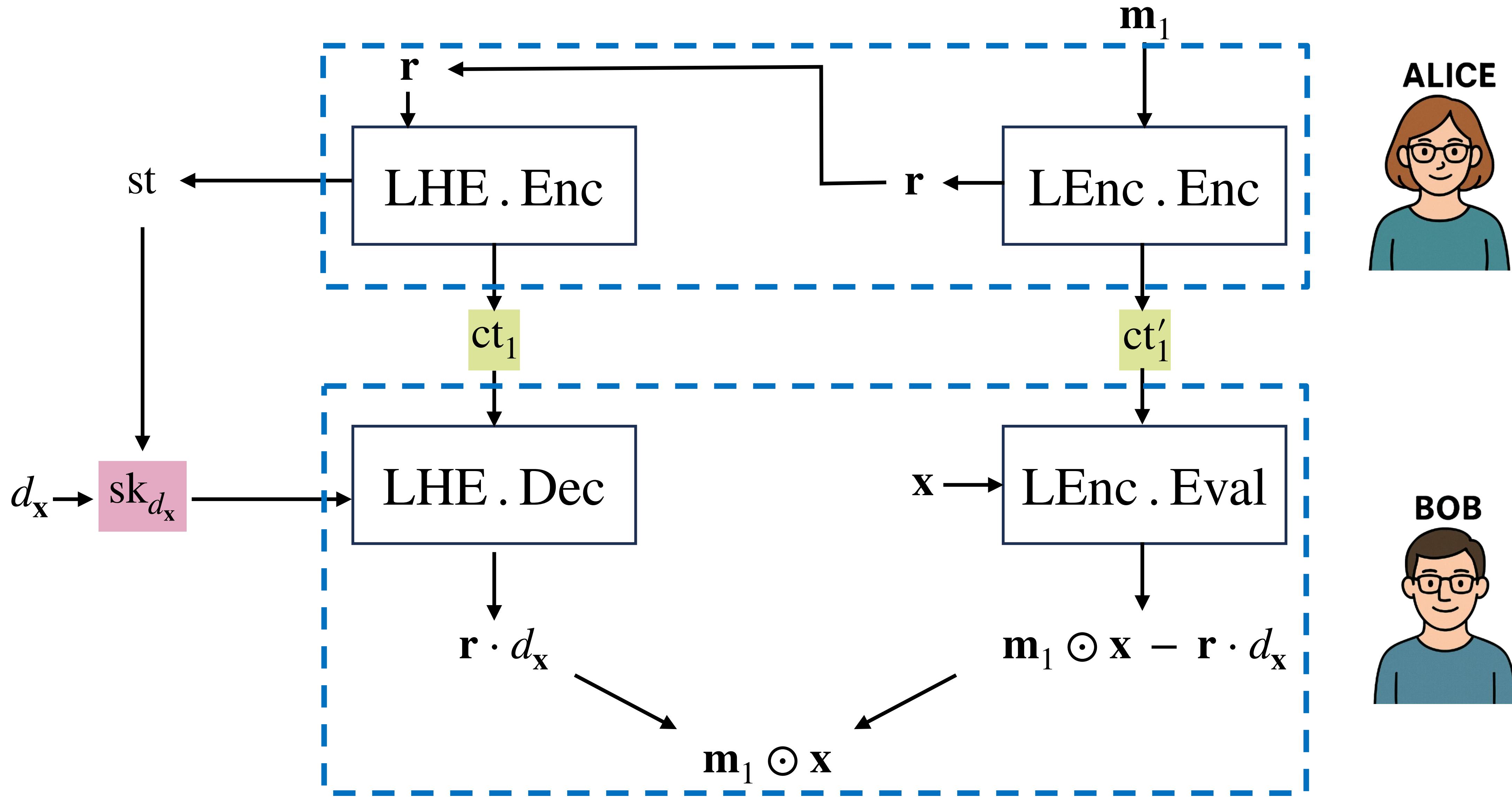
Constructing Batch-Select from RingLWE



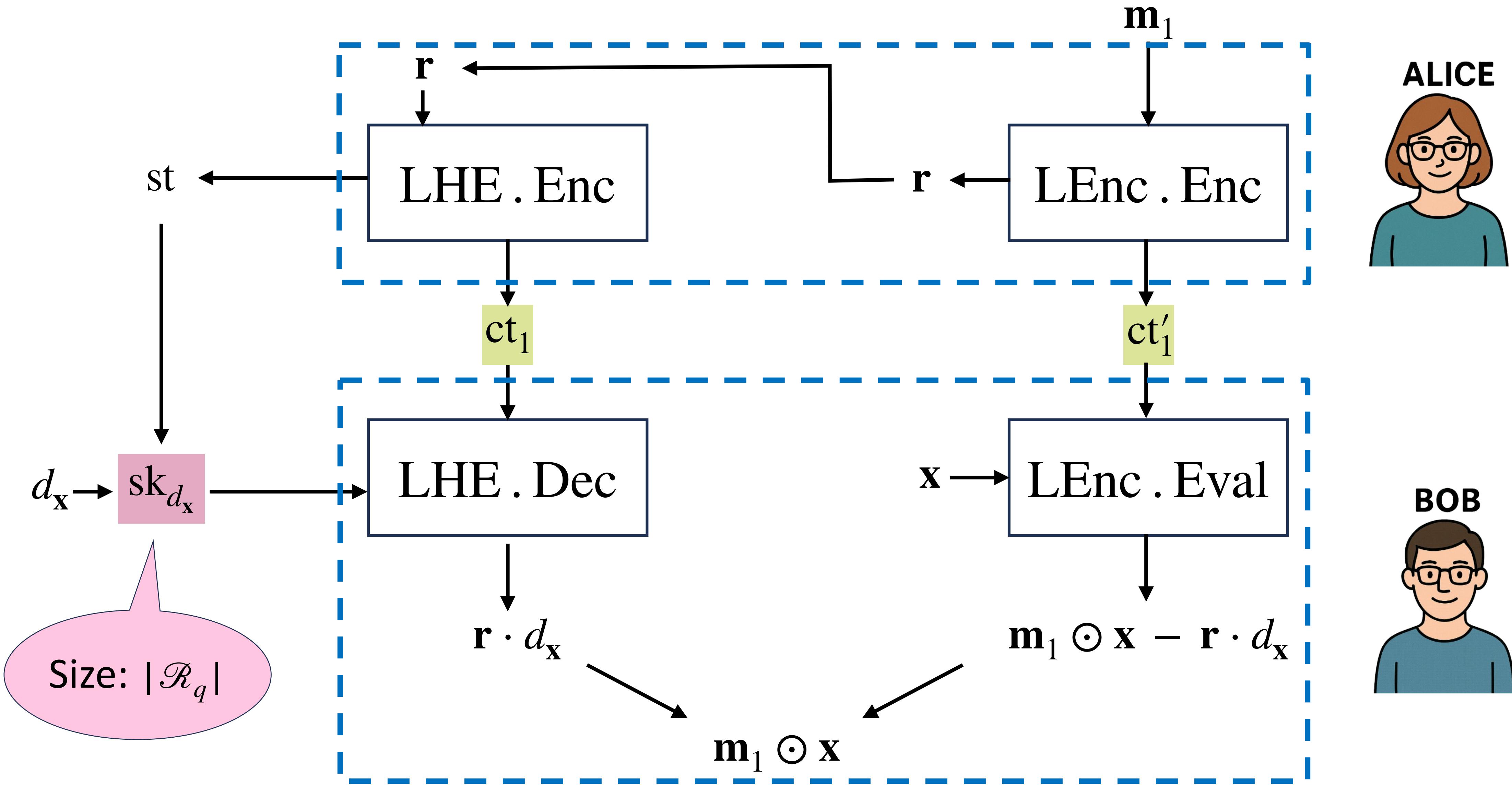
Constructing Batch-Select from RingLWE



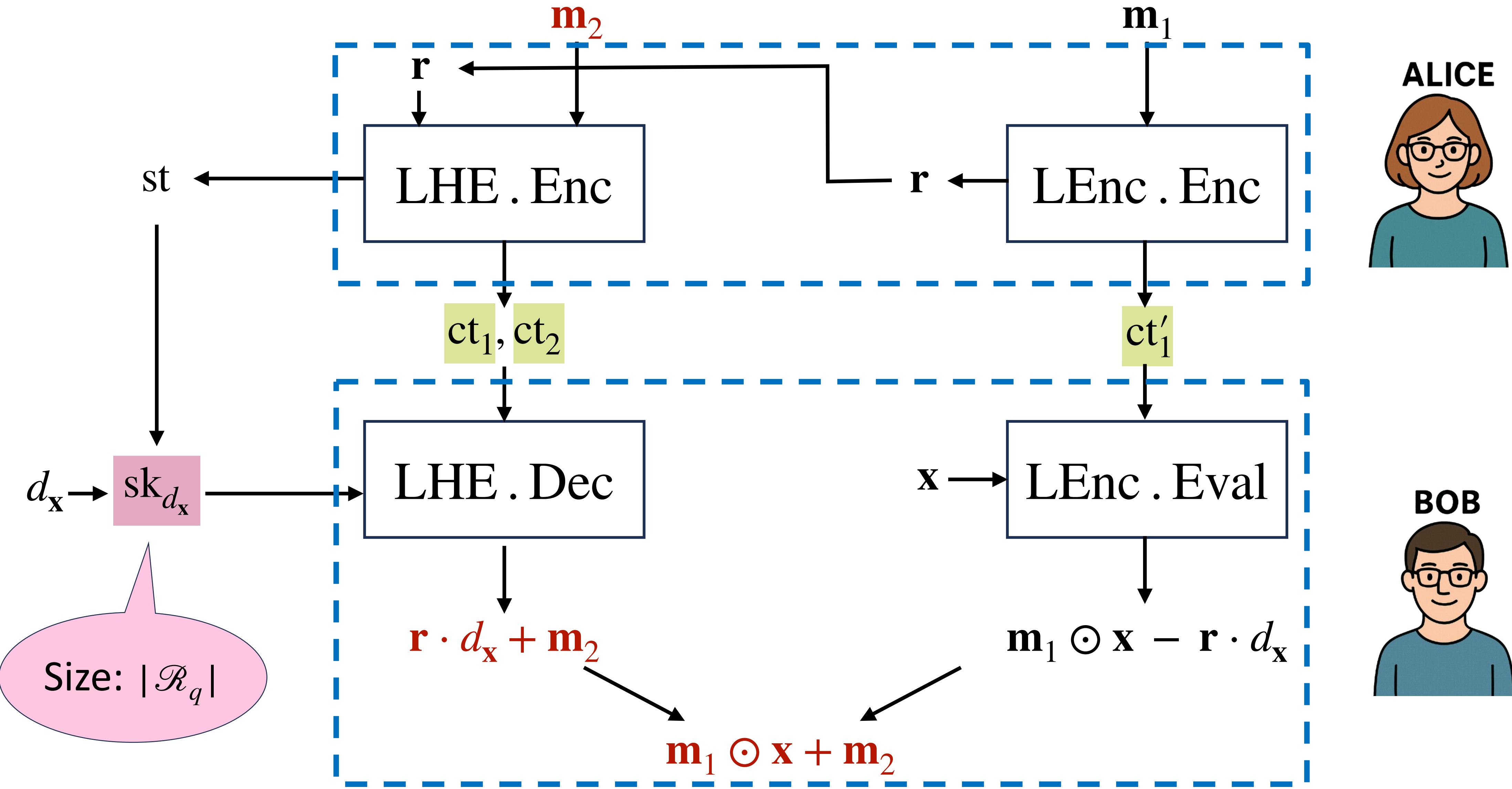
Constructing Batch-Select from RingLWE



Constructing Batch-Select from RingLWE



Constructing Batch-Select from RingLWE



Open Questions

- *Practical* label compression from other assumptions?
- Can succinct garbling *without offline phase* be made practical?

Open Questions

- *Practical* label compression from other assumptions?
- Can succinct garbling *without offline phase* be made practical?

Thank you!

<https://ia.cr/2024/2048>

References

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- [GS18]: Garg, S., & Srinivasan, A. (2018). **Adaptively secure garbling with near optimal online complexity**. In *Advances in Cryptology–EUROCRYPT 2018: 37th Annual International Conference on the Theory and Applications of Cryptographic Techniques, Tel Aviv, Israel, April 29-May 3, 2018 Proceedings, Part II 37* (pp. 535-565). Springer International Publishing.
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- Images generated with ChatGPT