

A General Framework for Registered Functional Encryption via User-Specific Pre-Constraining

Tapas Pal¹

Robert Schädlich²

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¹ Karlsruhe Institute of Technology, KASTEL Security Research Labs

² DIENS, École normale supérieure, PSL University, CNRS, Inria

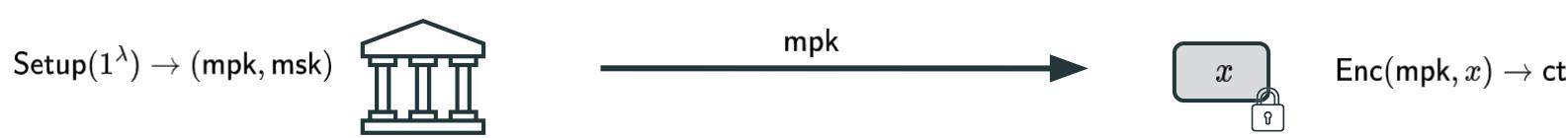


Functional Encryption (FE) [TCC:BSW11]

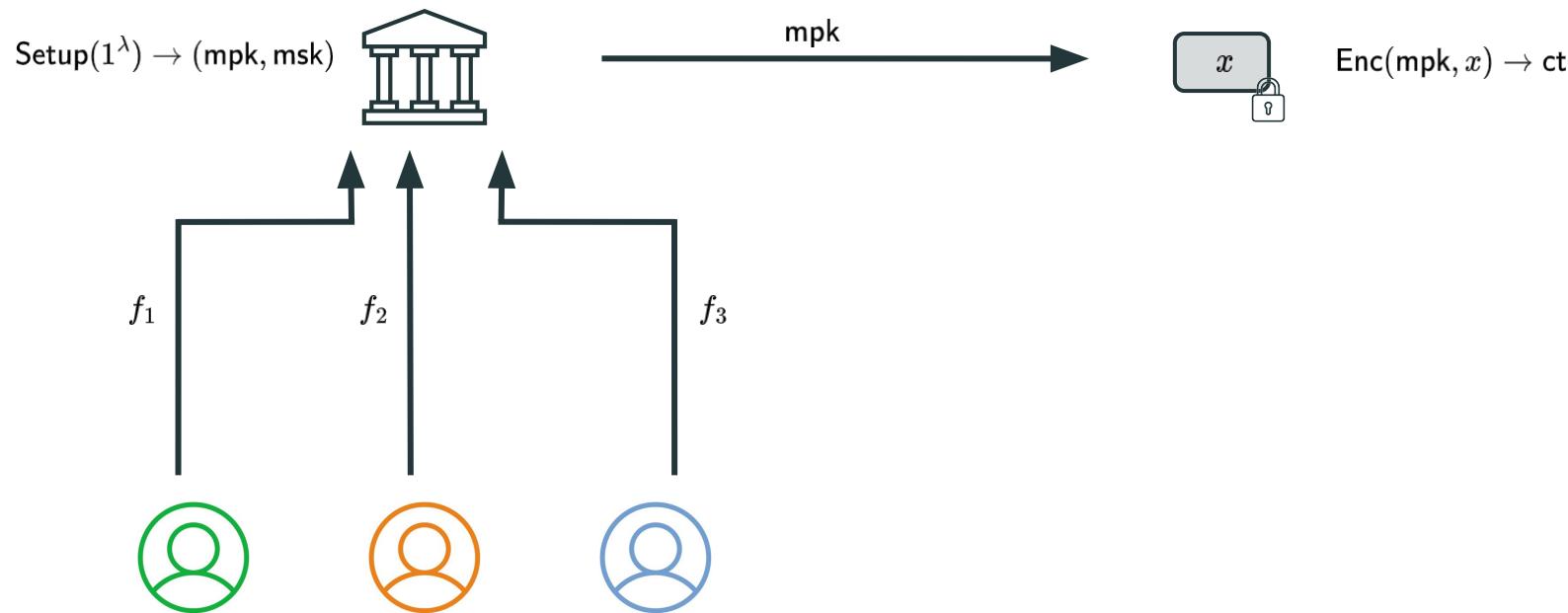
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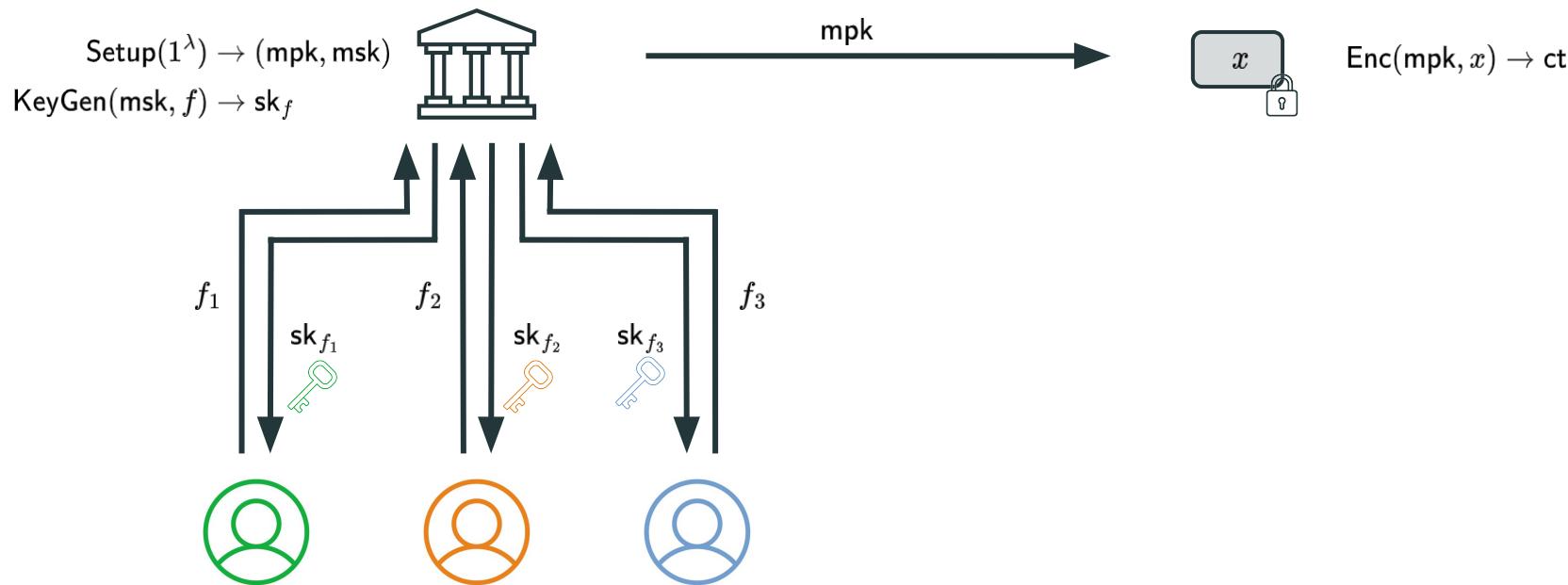
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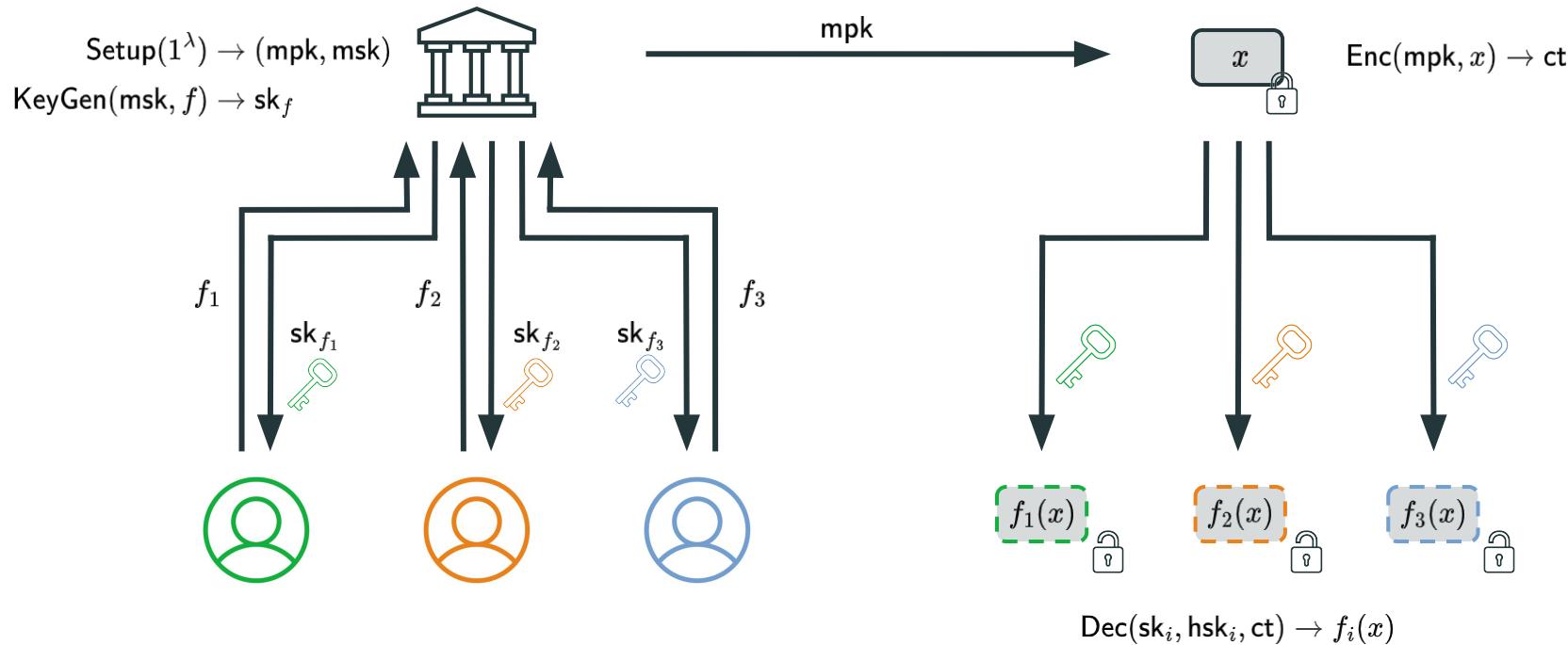
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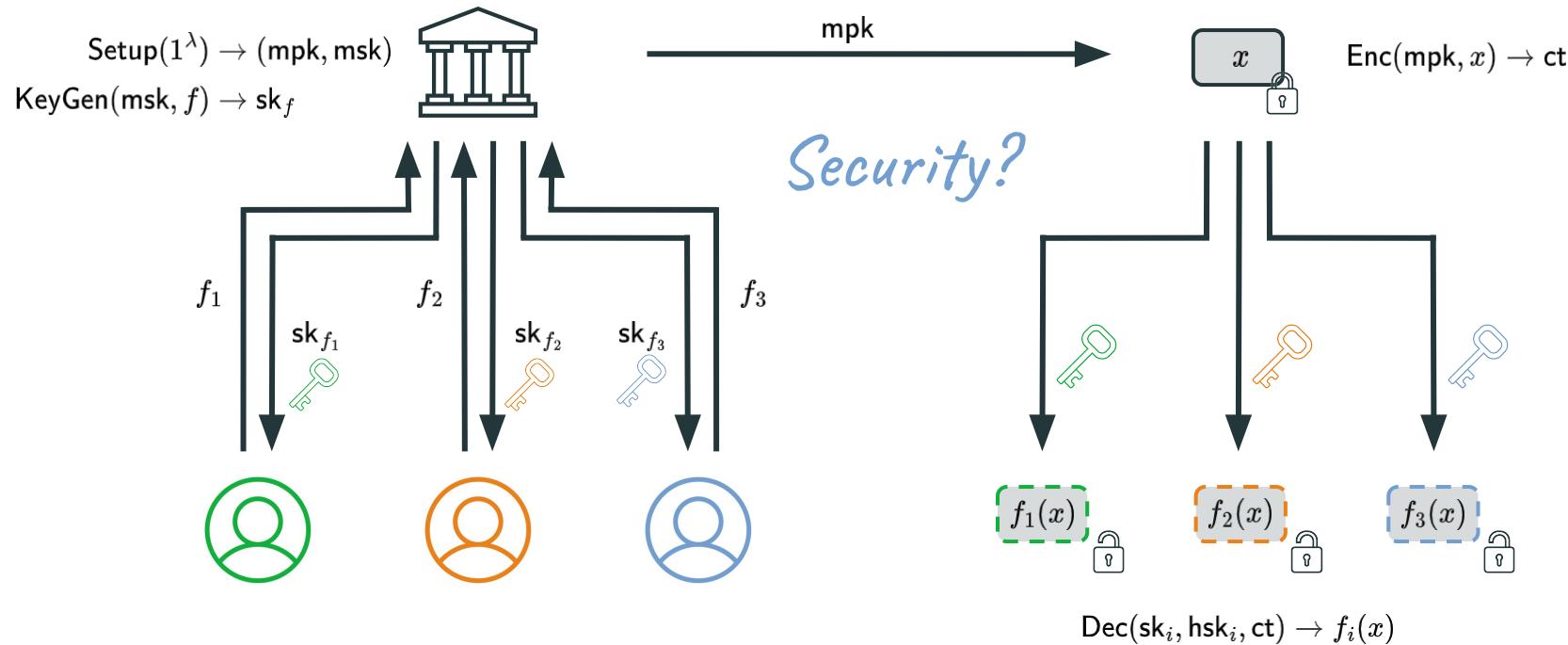
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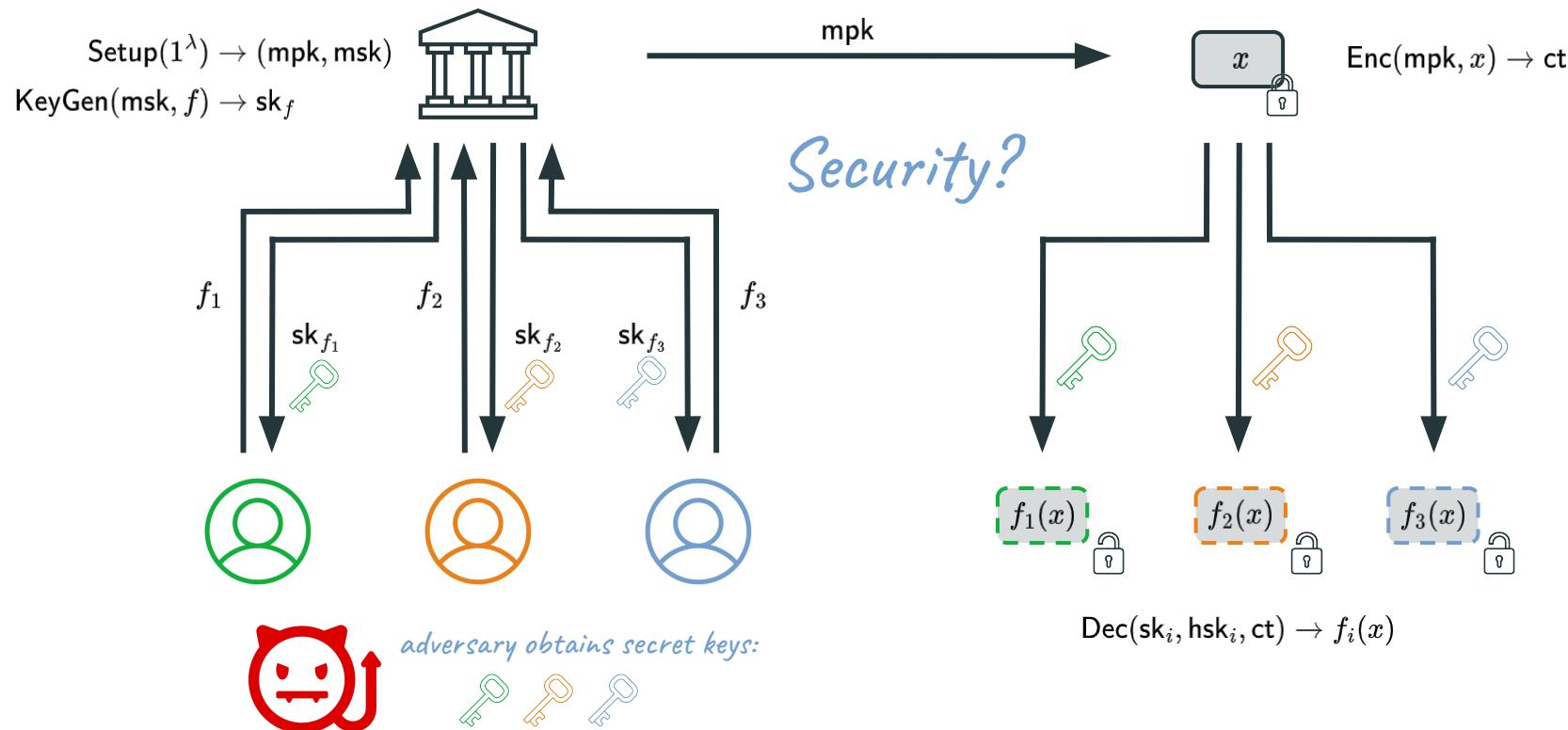
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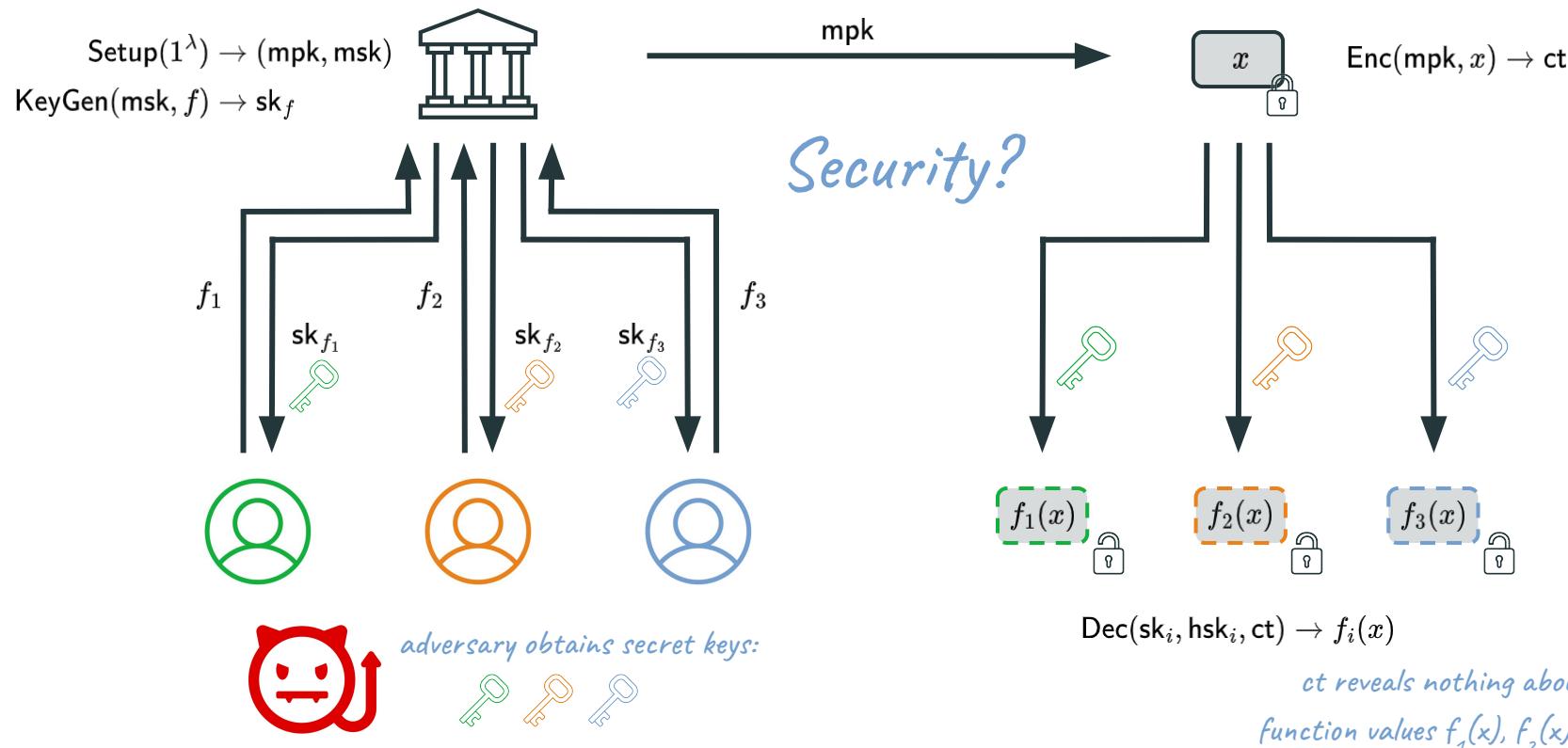
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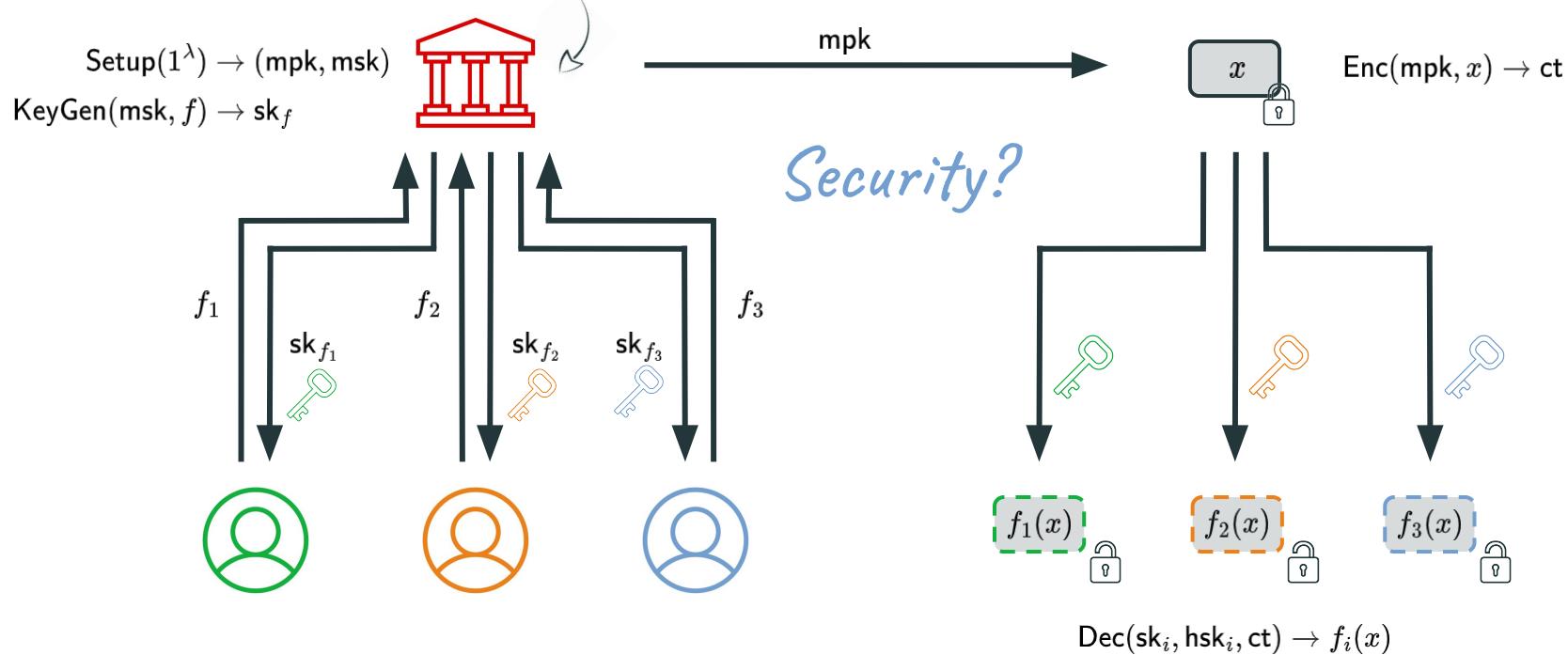


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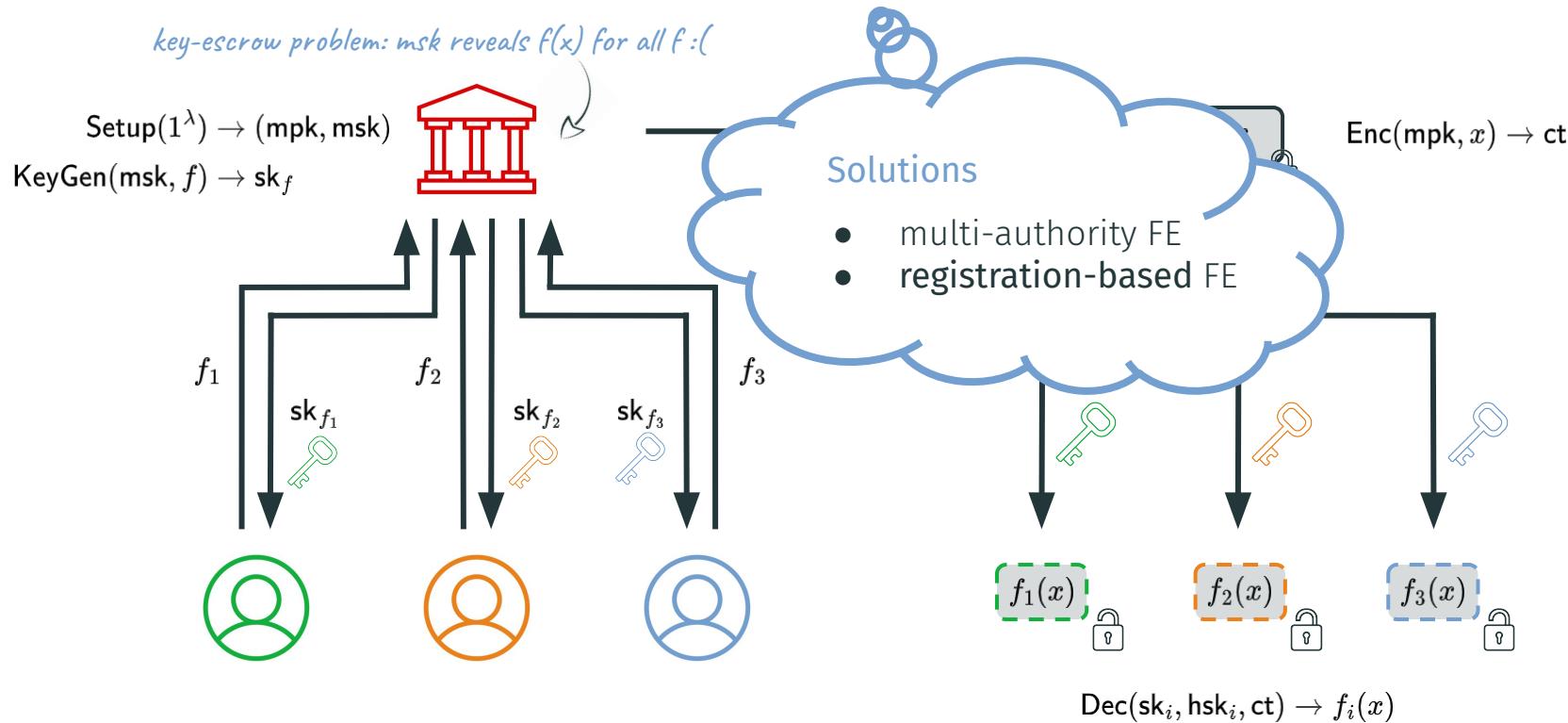


Functional Encryption (FE) [TCC:BSW11]

key-escrow problem: msk reveals $f(x)$ for all f :



Functional Encryption (FE) [TCC:BSW11]



Registered Functional Encryption (RFE) [AC:FFM+23]

$\text{Setup}(1^\lambda) \rightarrow \text{crs}$



pk_1, sk_1



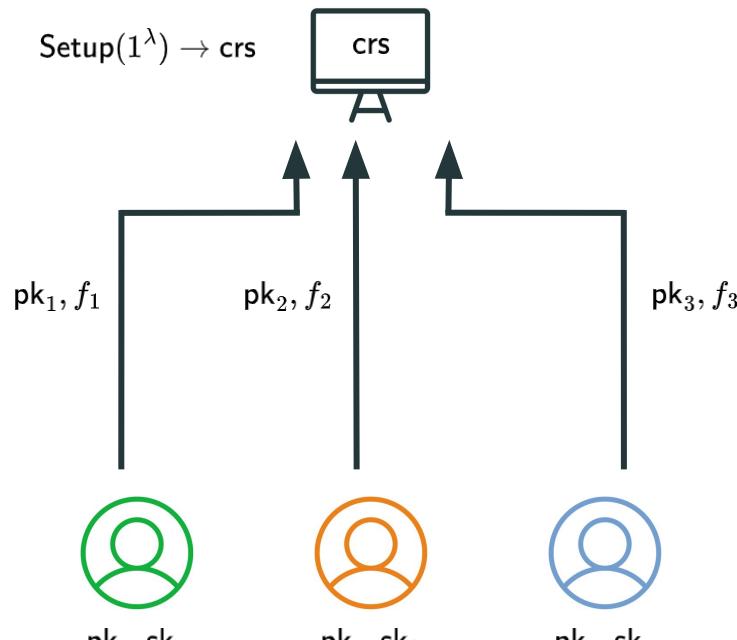
pk_2, sk_2



pk_3, sk_3

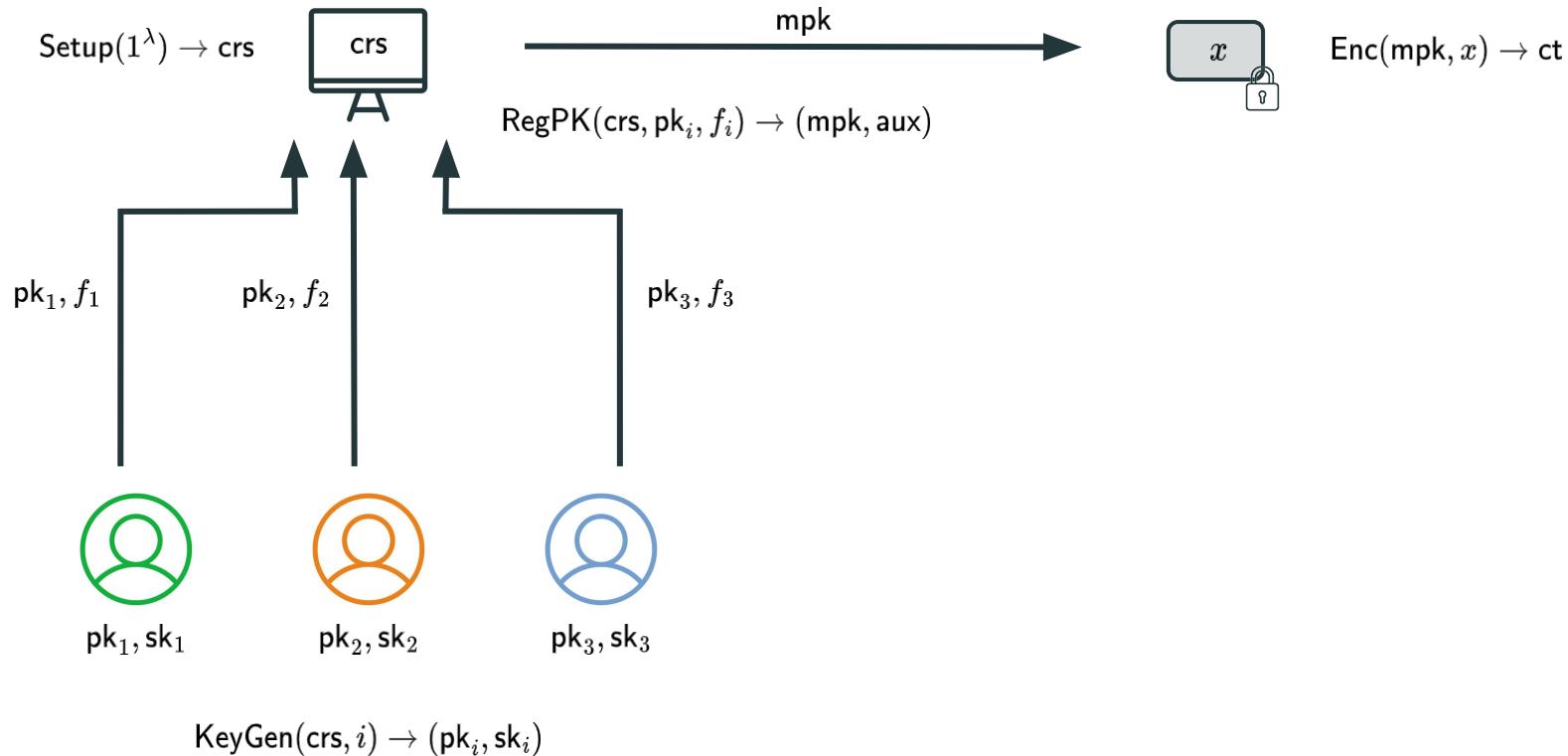
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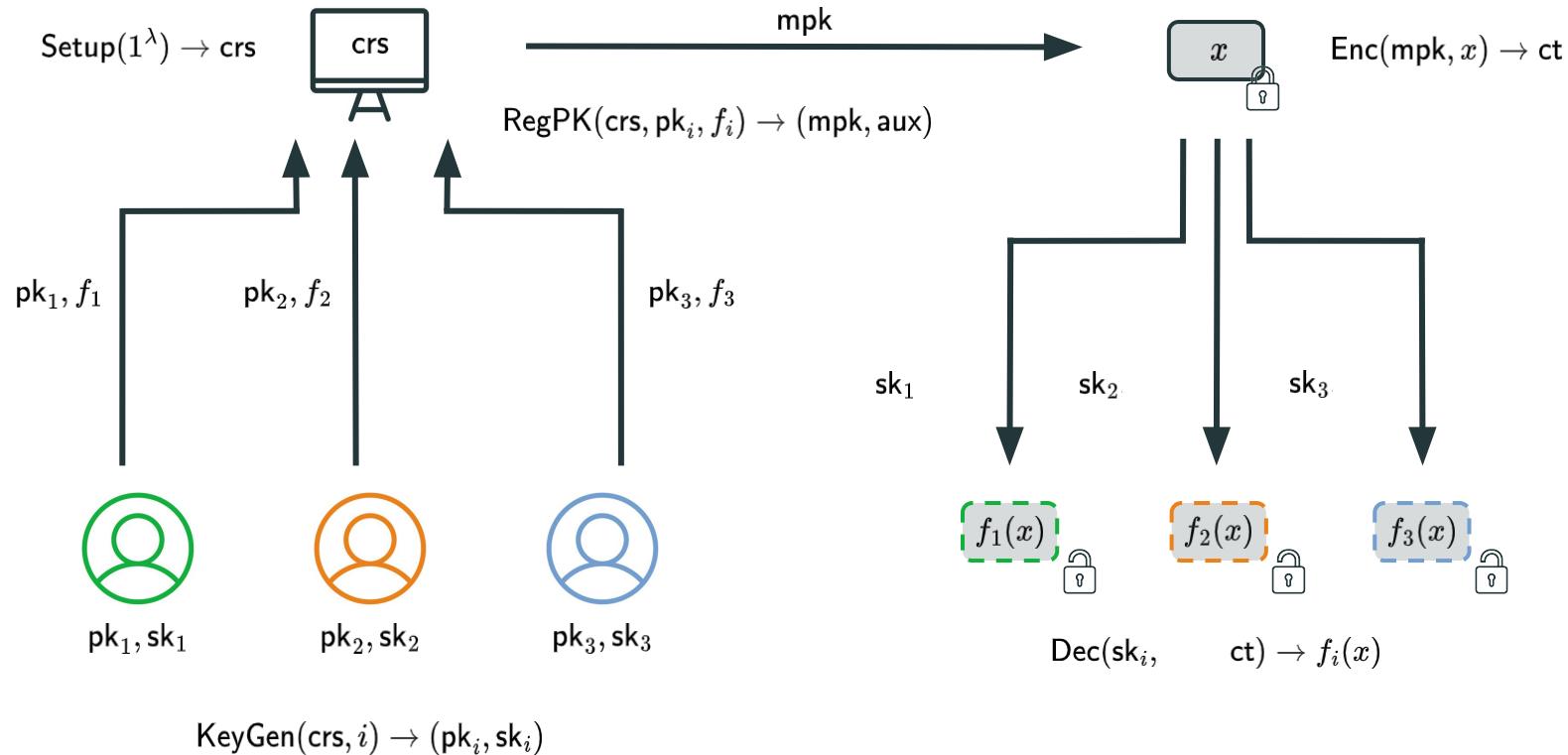


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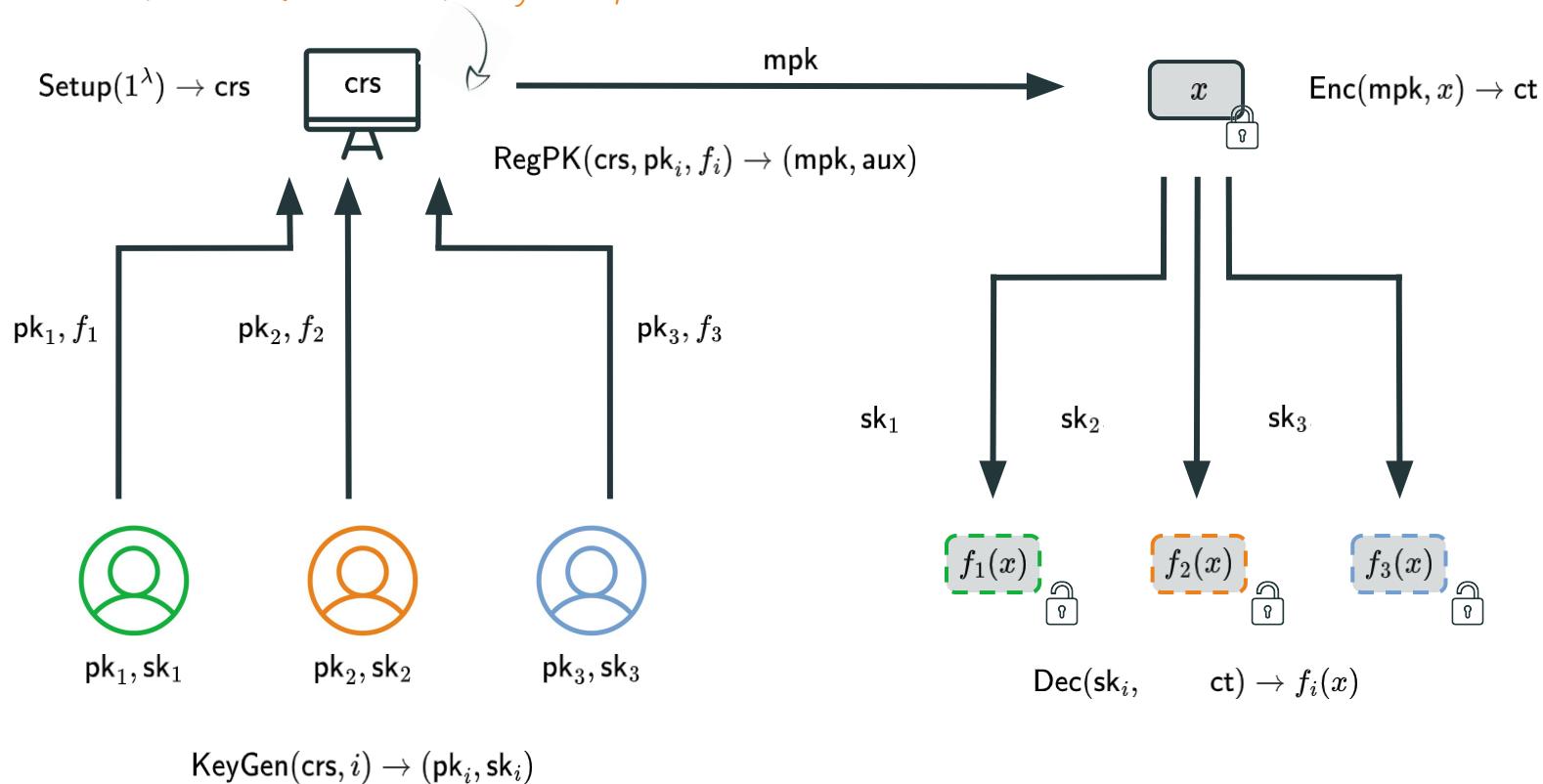


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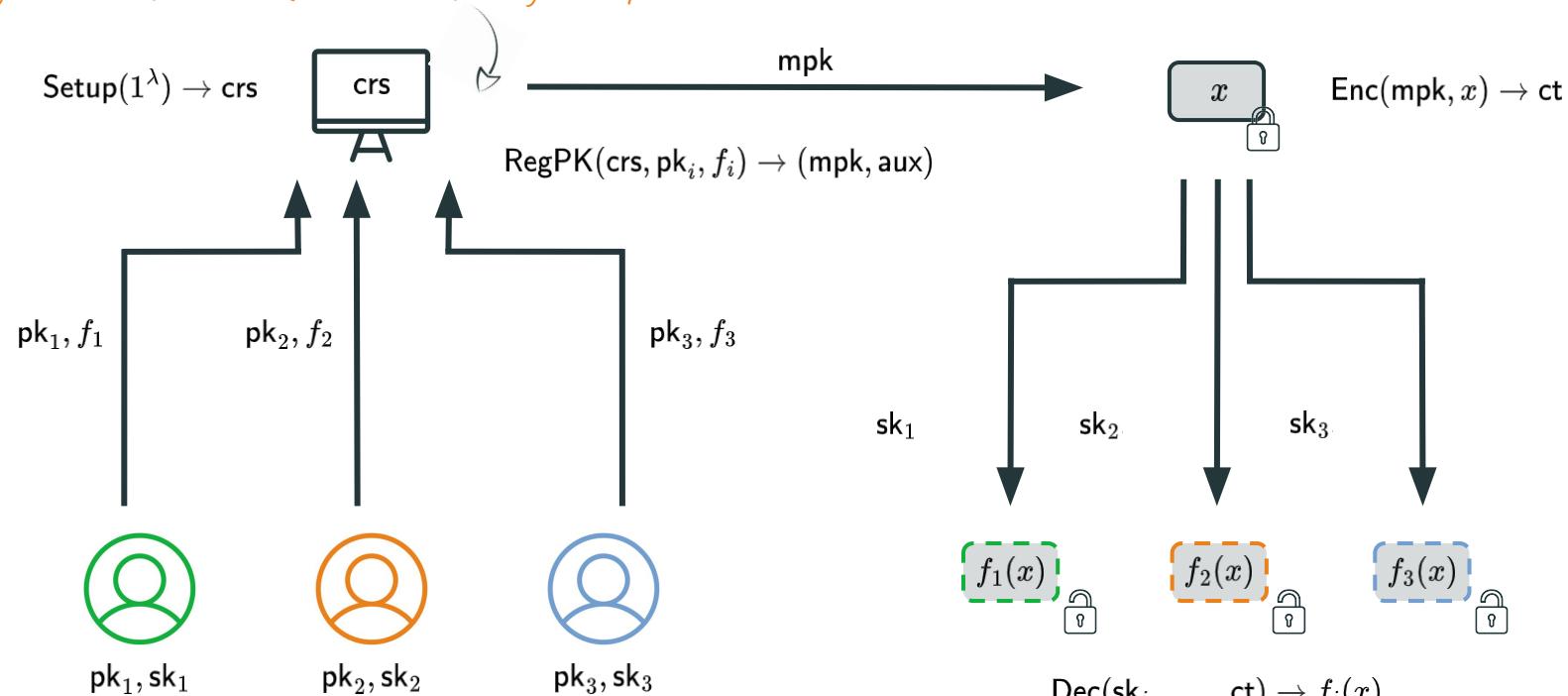
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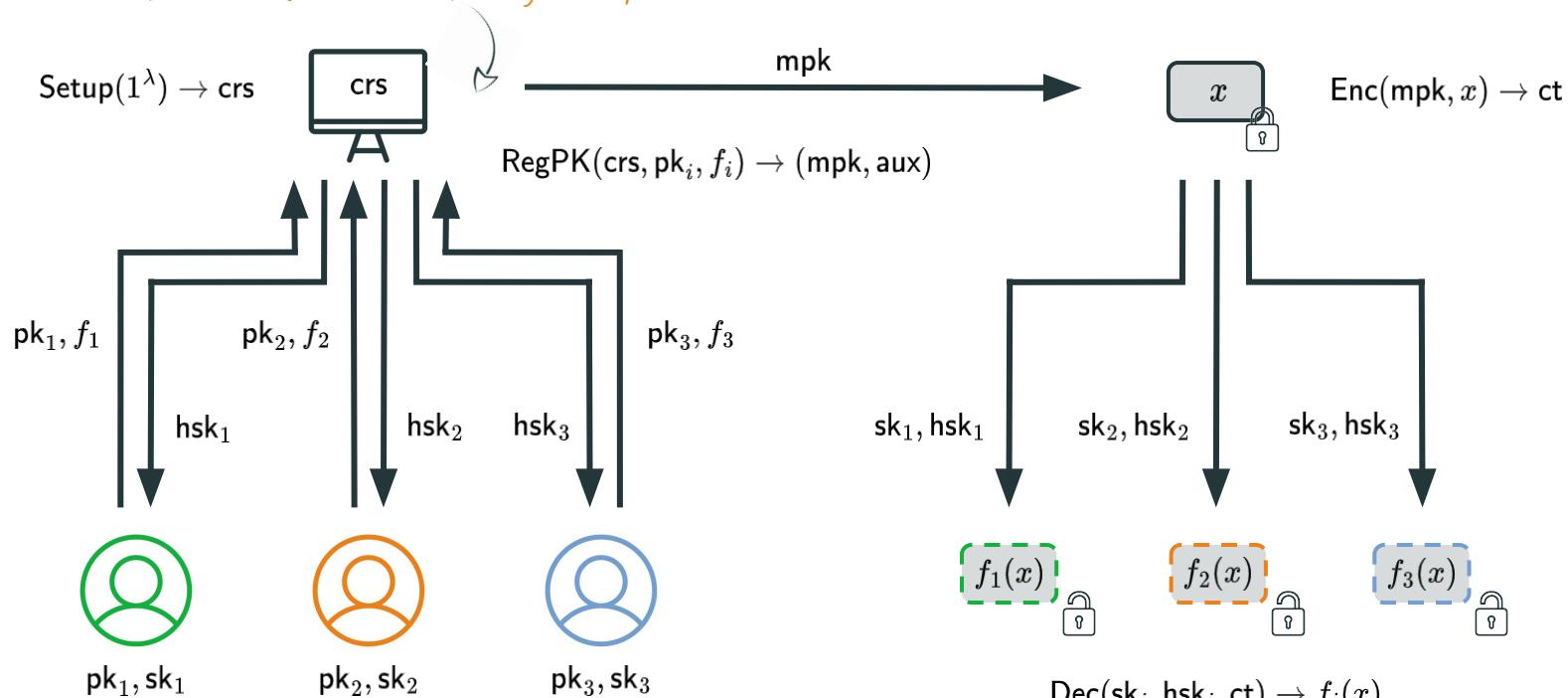


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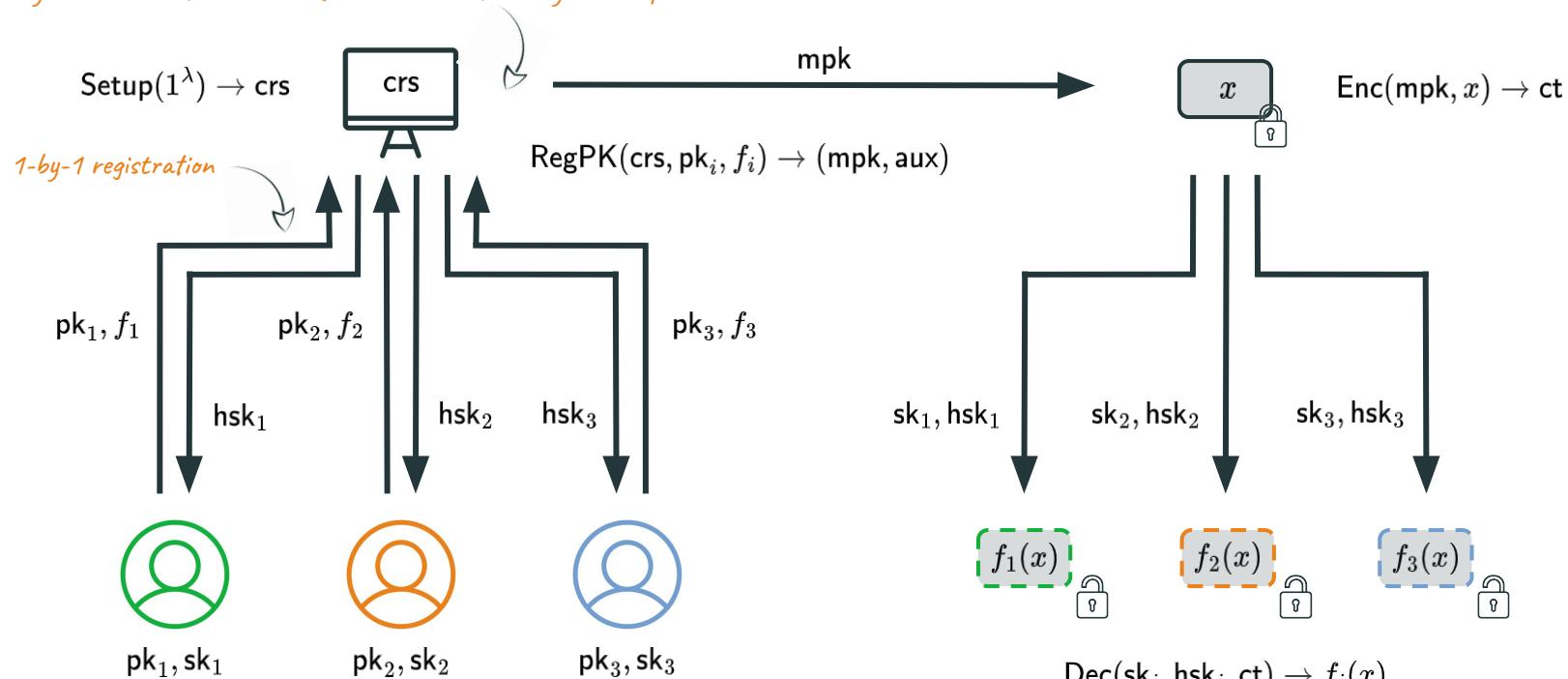


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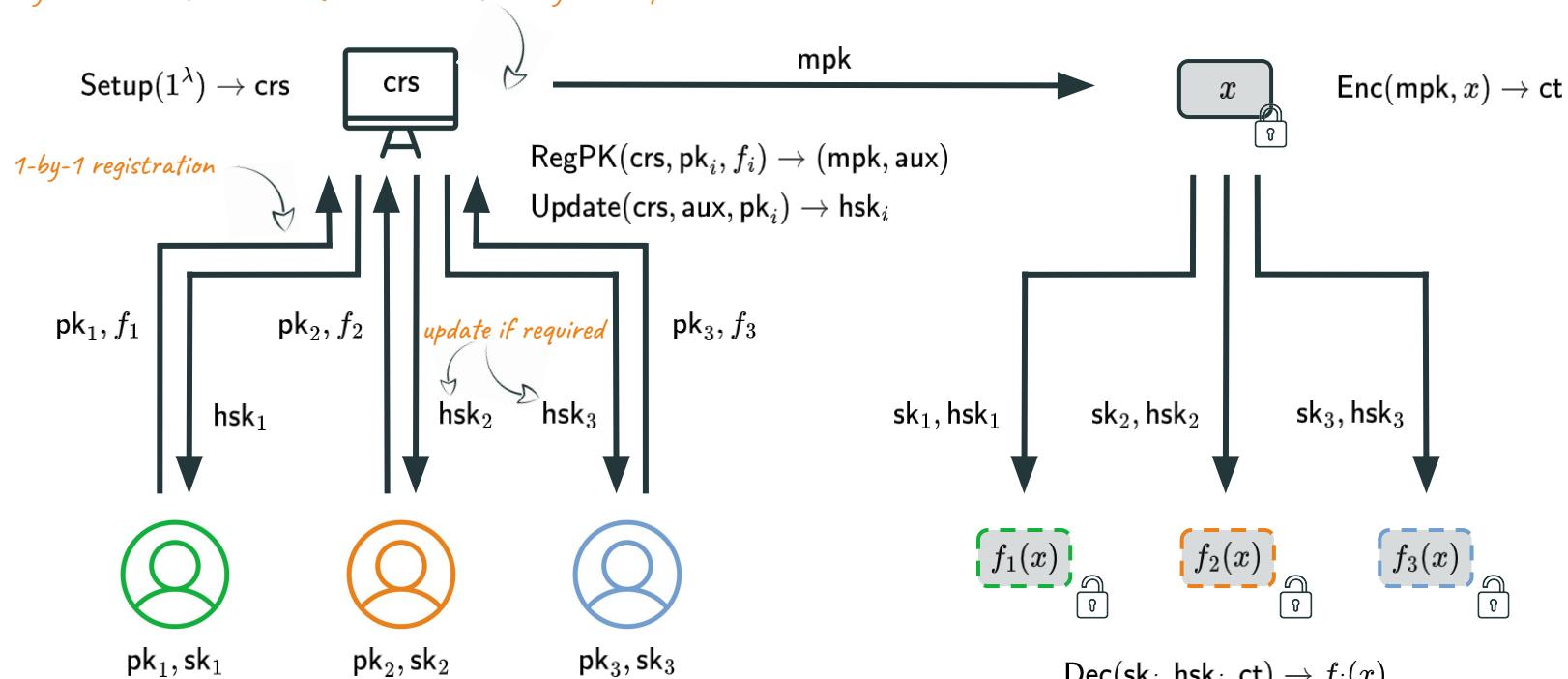


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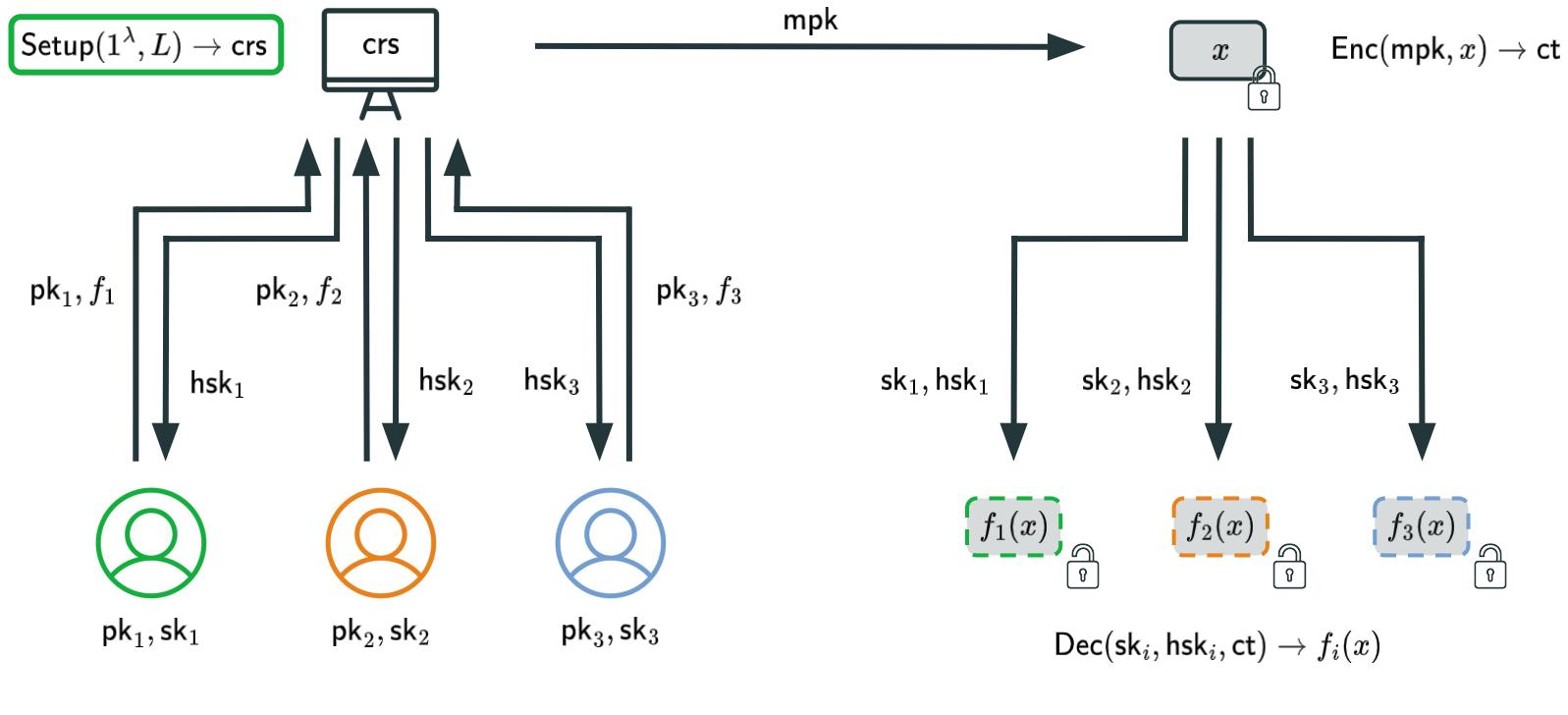
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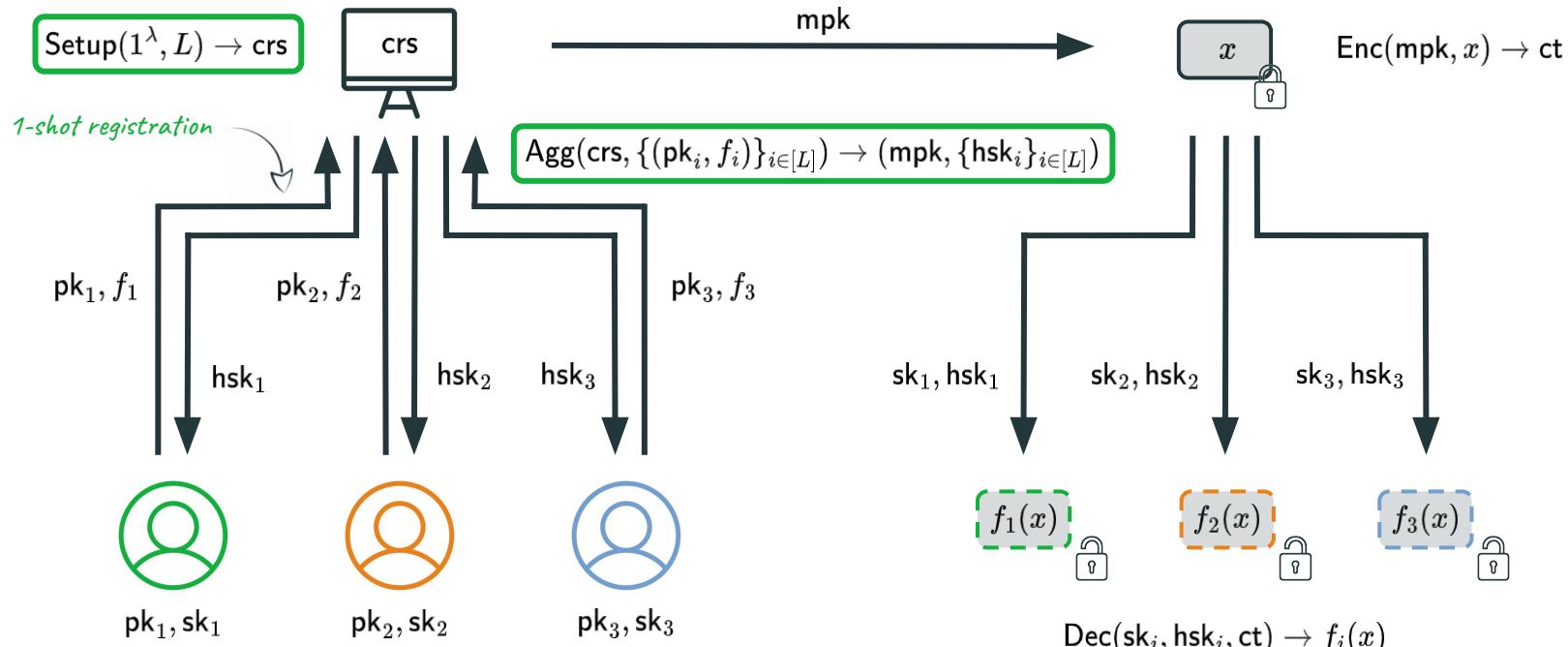
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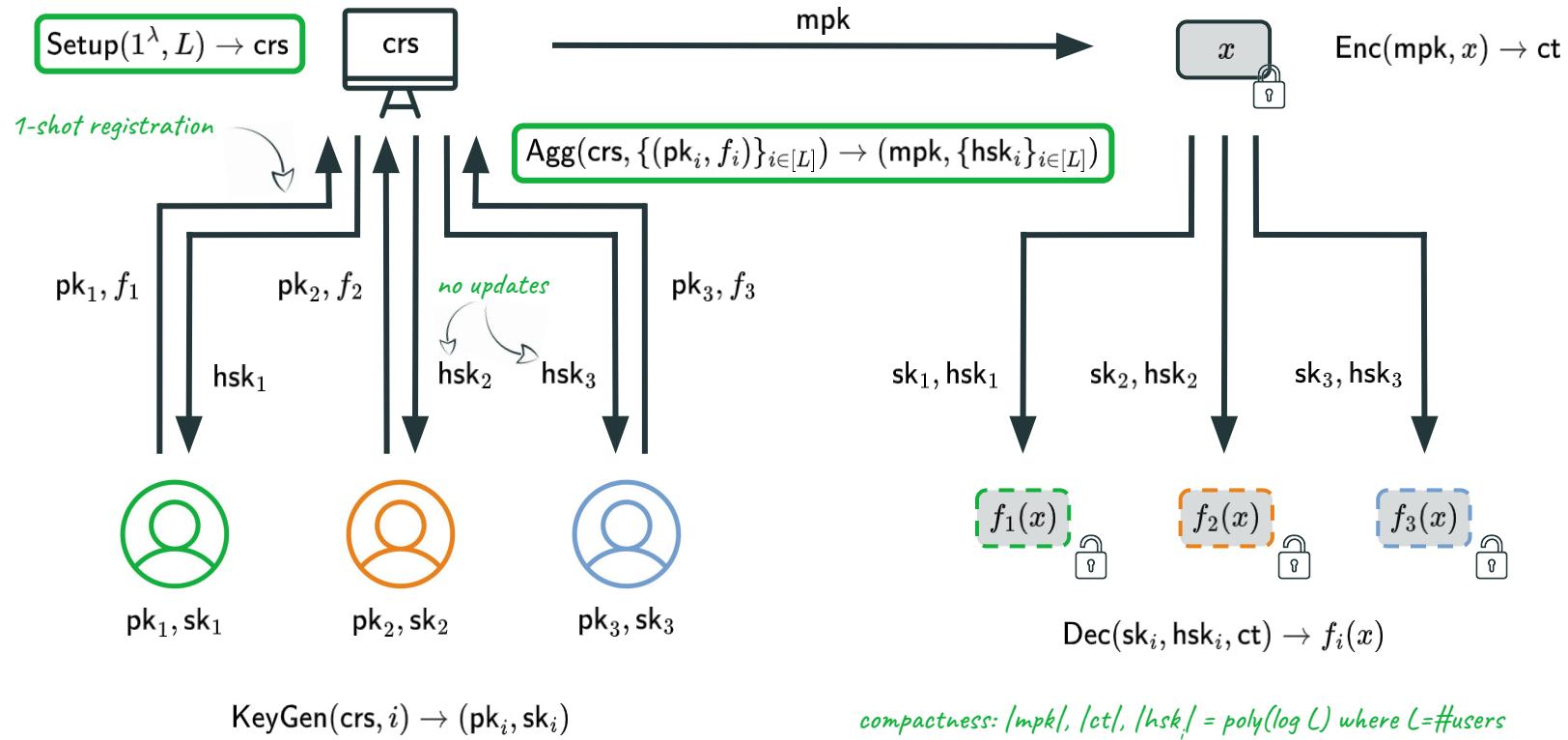
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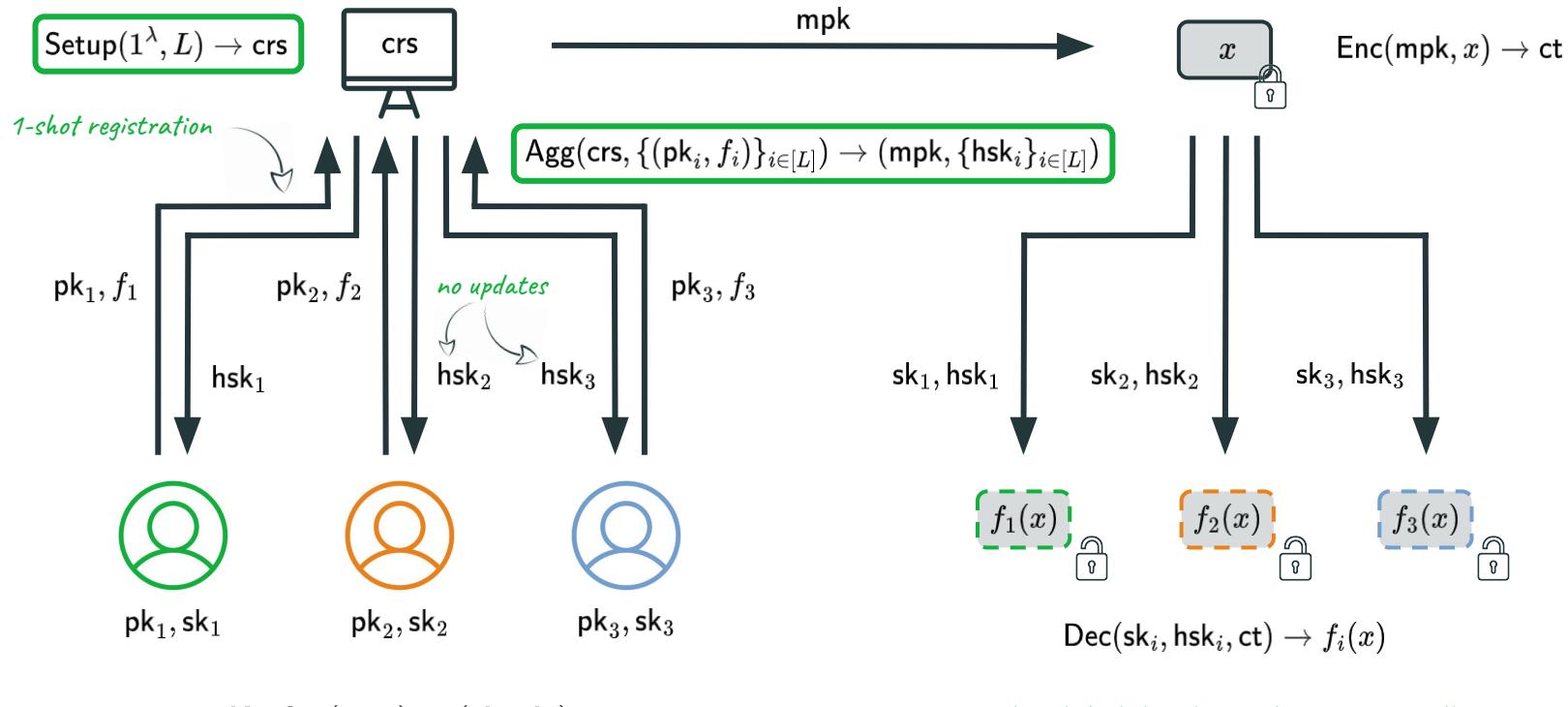


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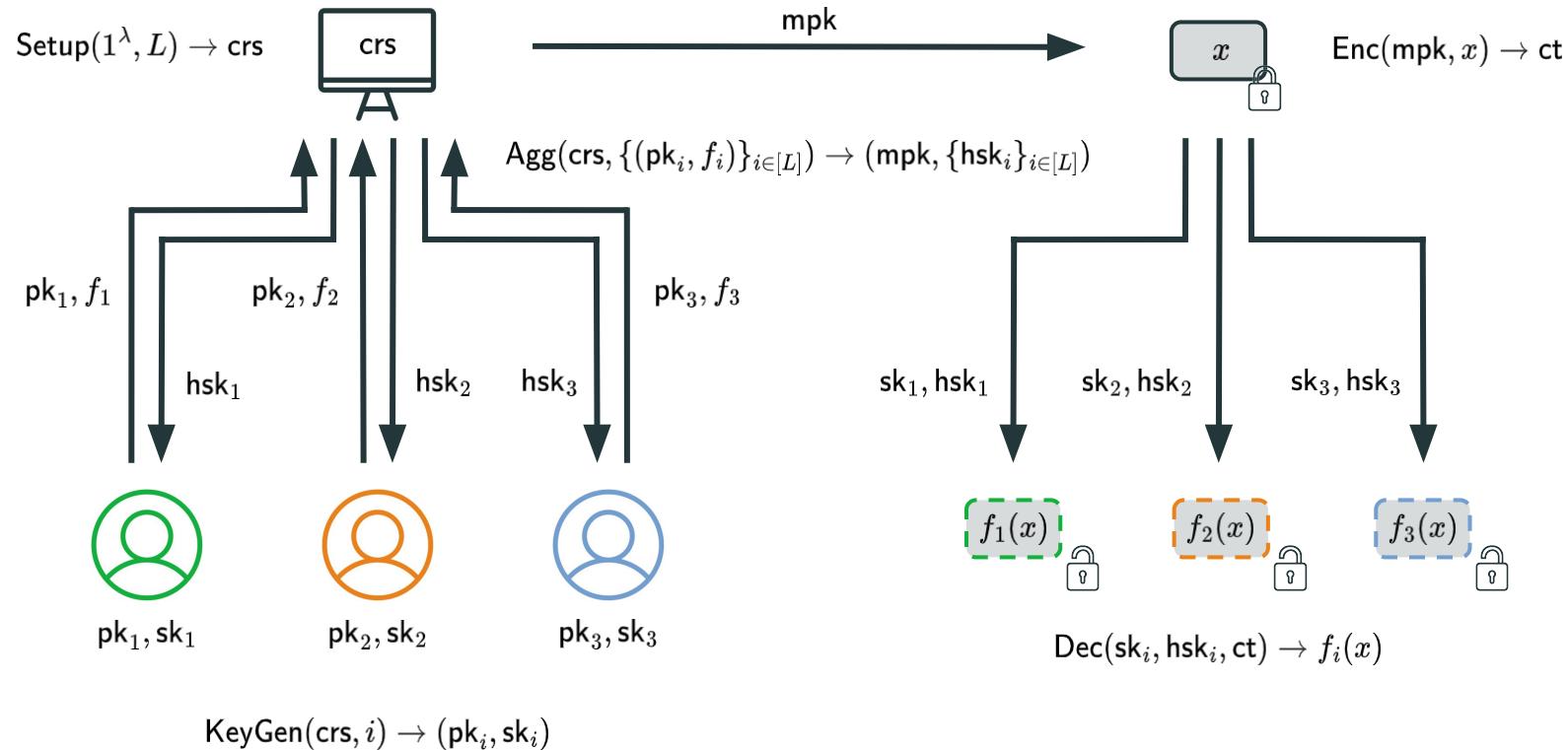


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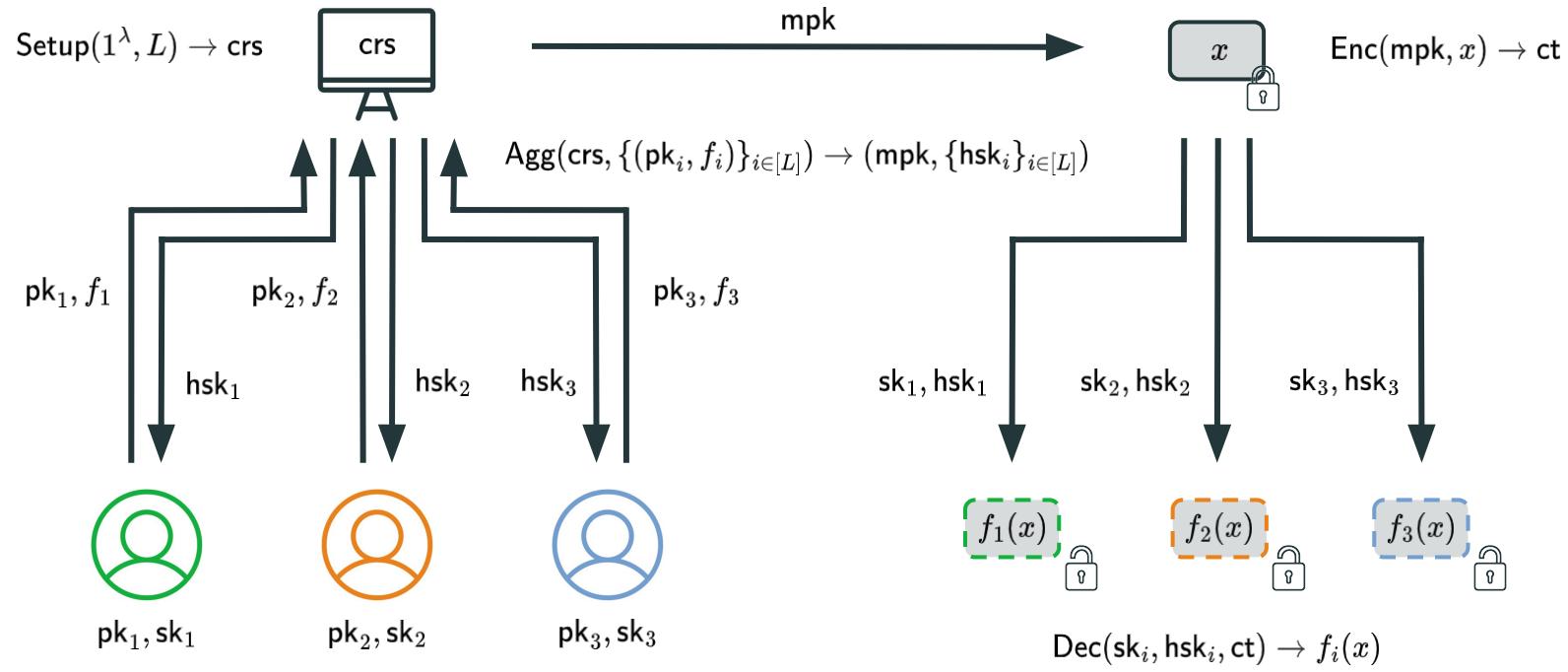
[HLWW23]: sRFE \Rightarrow RFE (“powers-of-two compiler”)



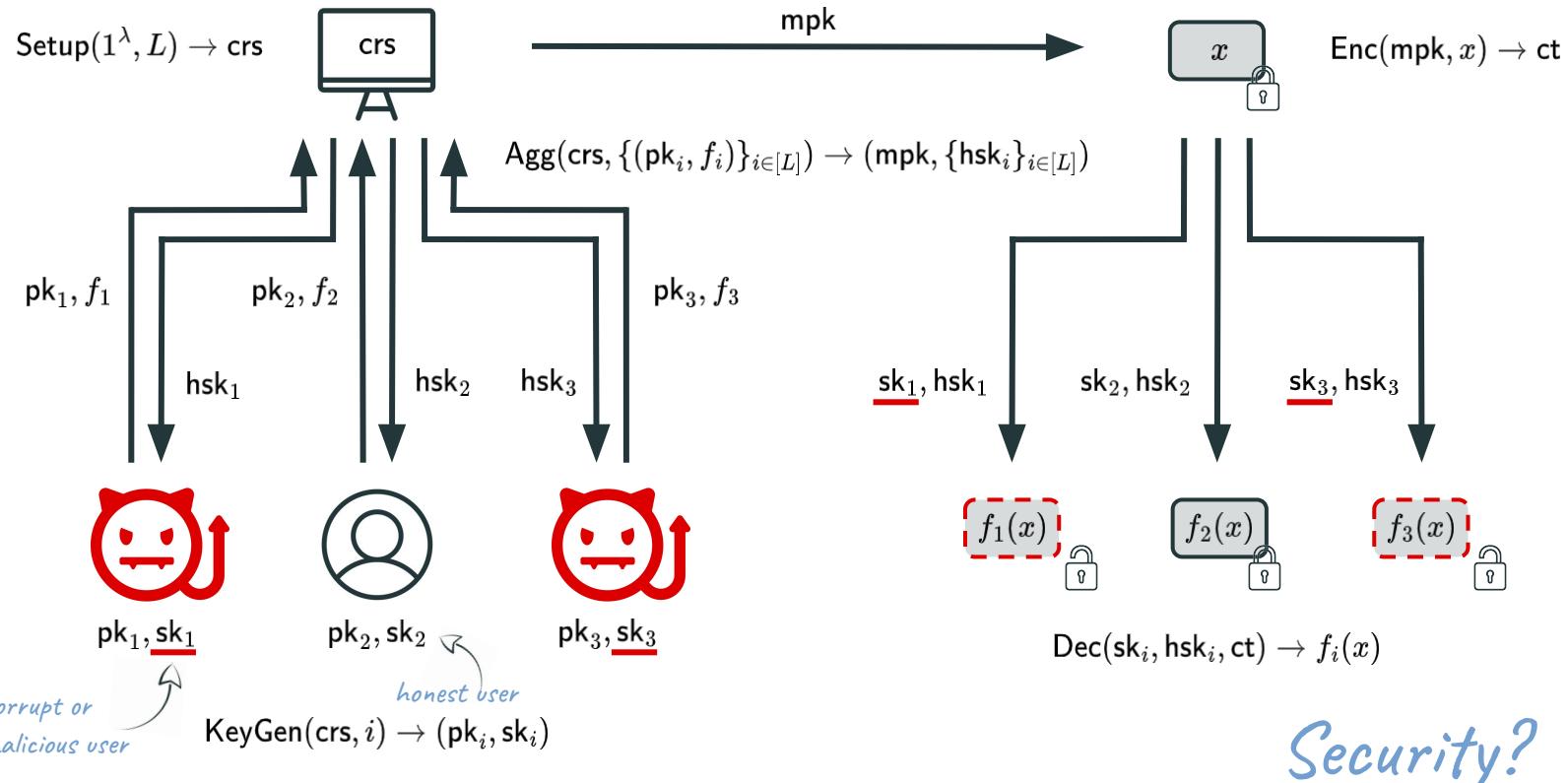
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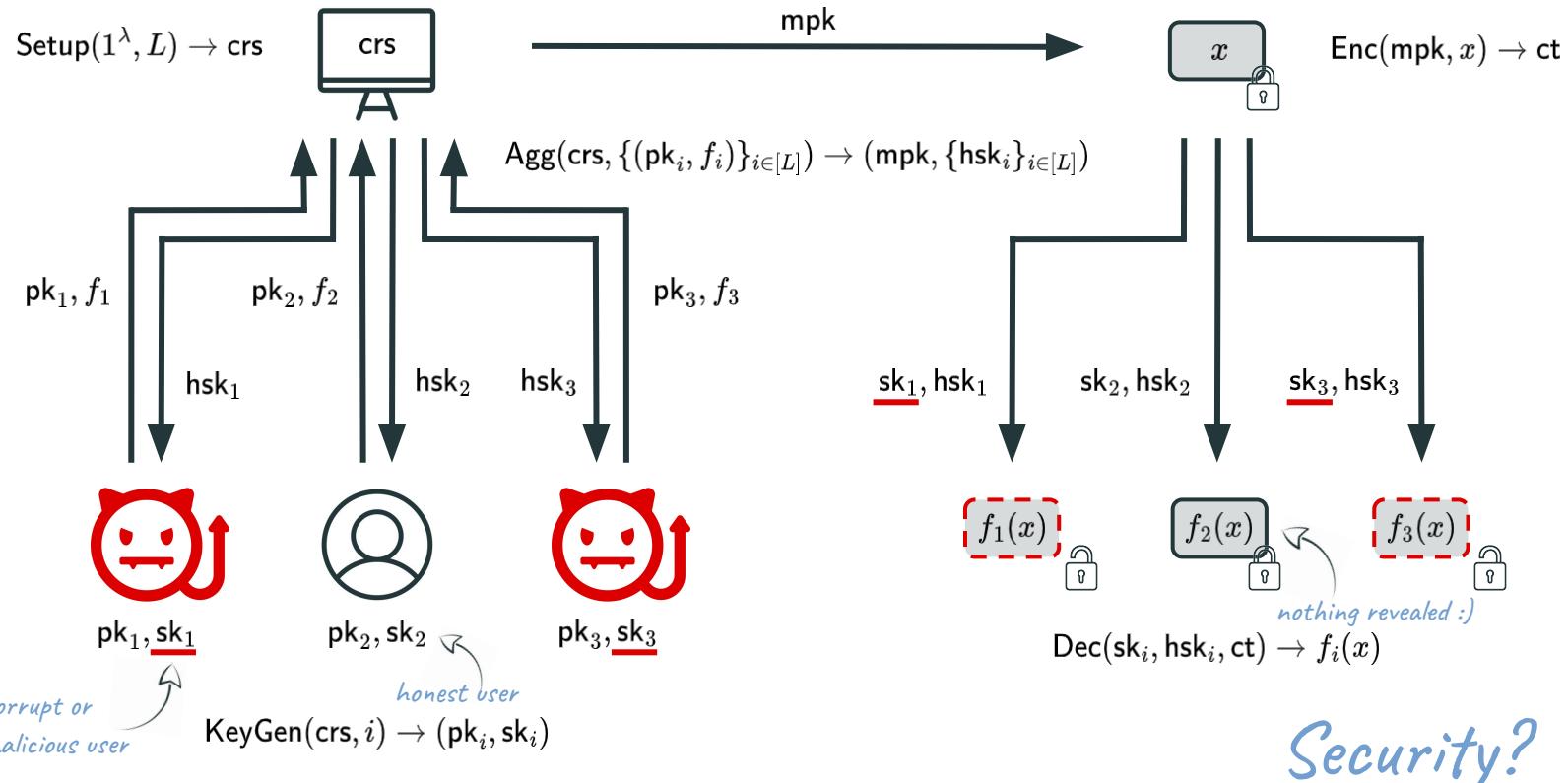
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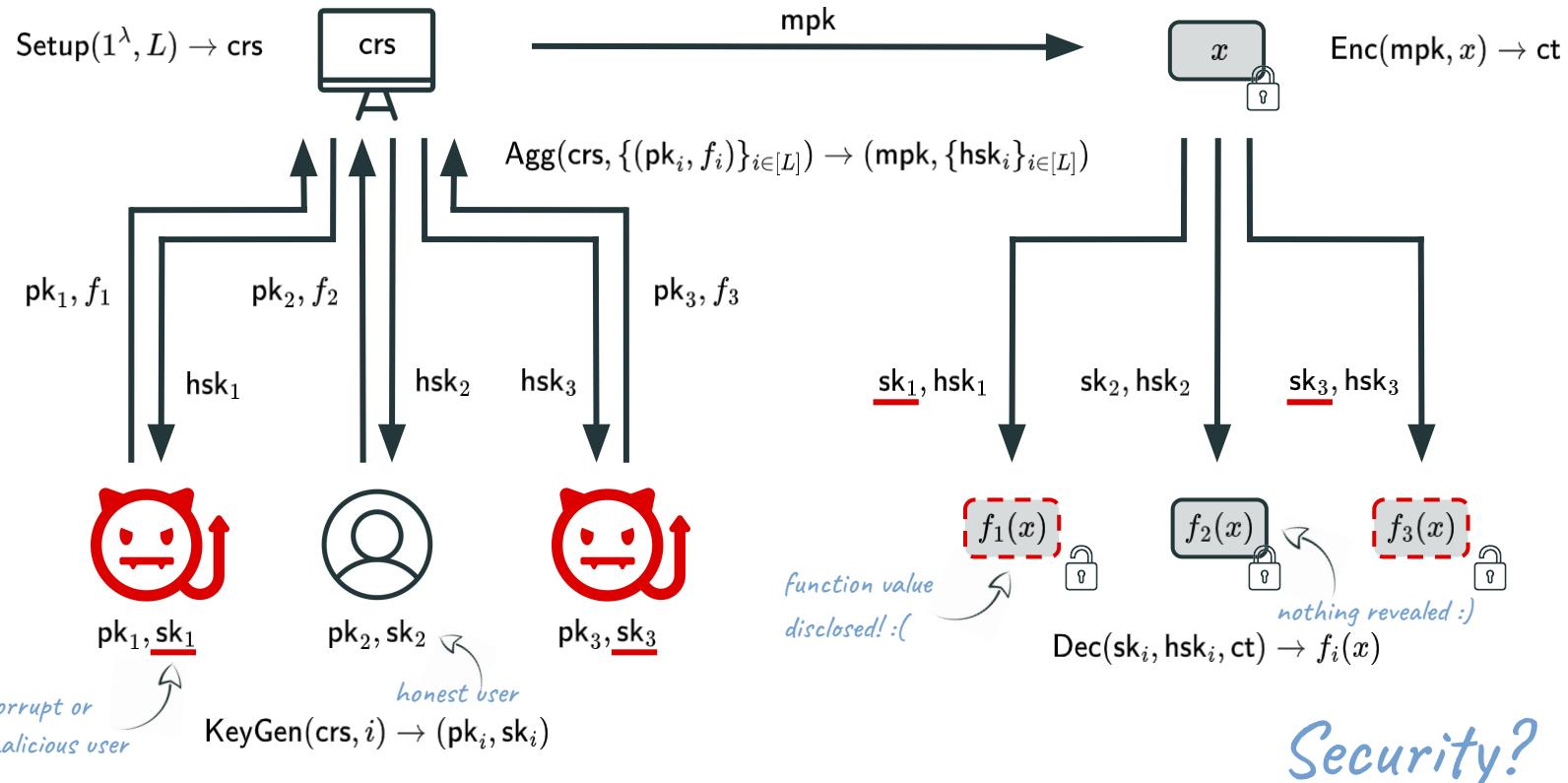
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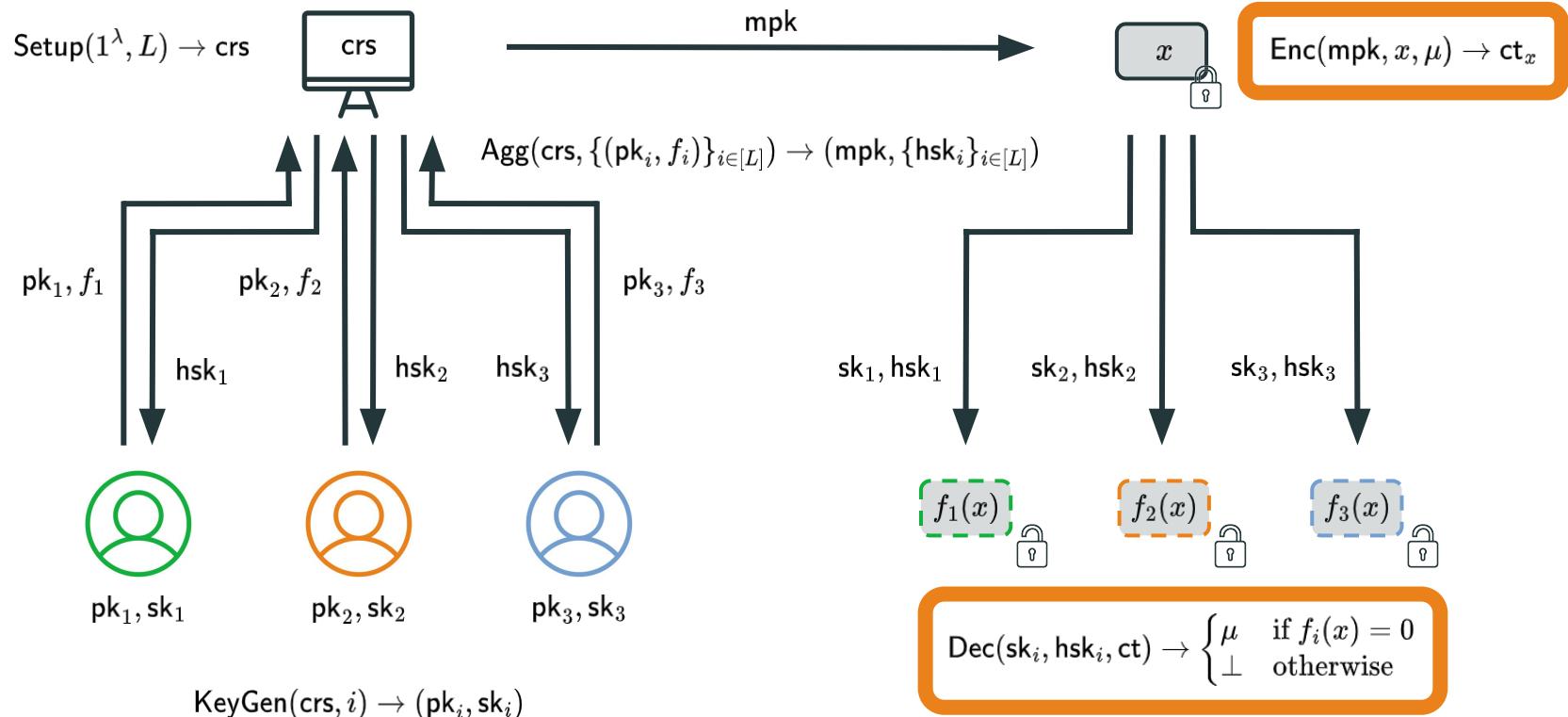
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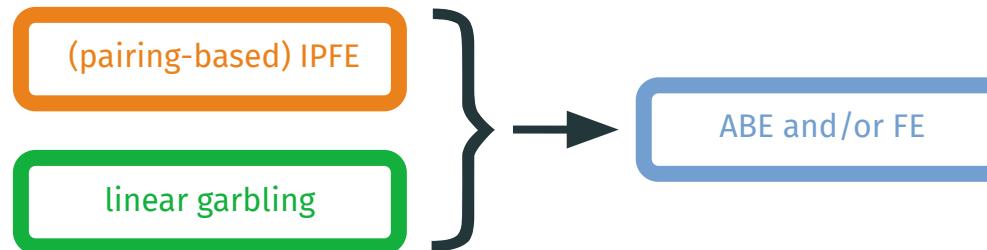
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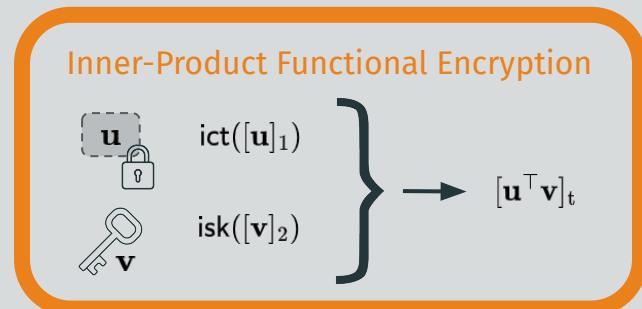
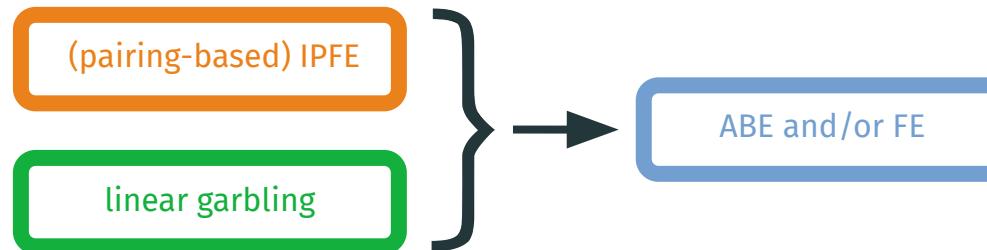
Special Case: Registered ABE



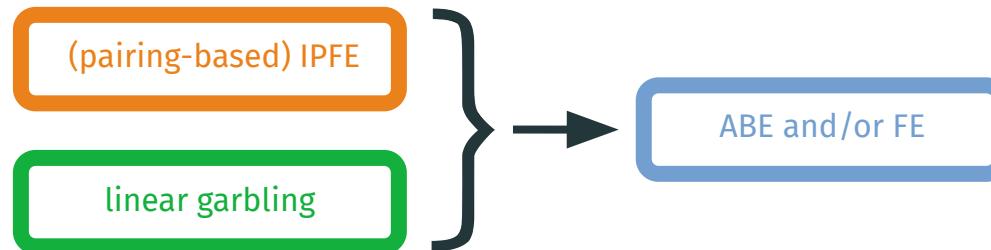
Framework for *Non-Registered* ABE



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Arithmetic Key Garbling Scheme [EC:LL20]

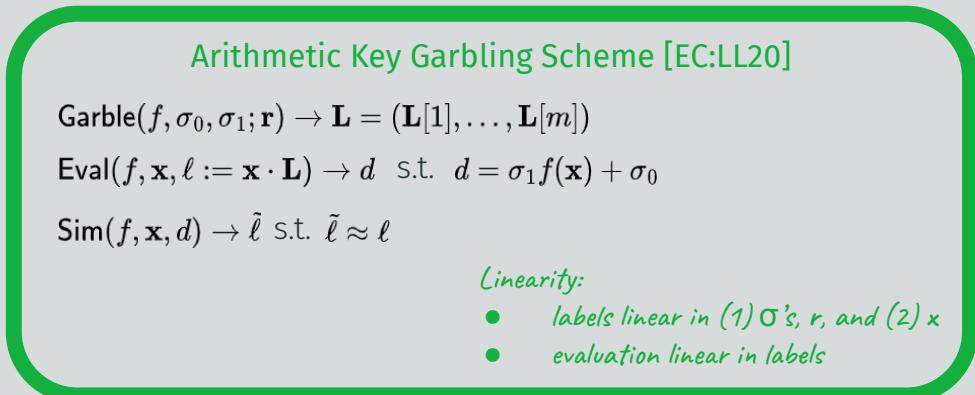
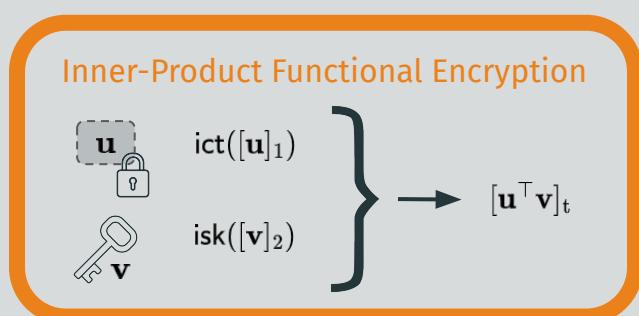
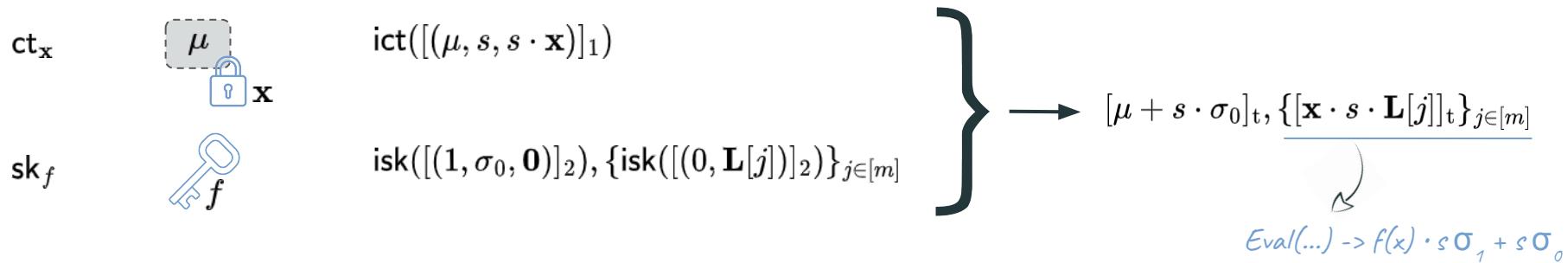
2-step garbling procedure:

1. given f and **secret** inputs σ_0, σ_1 , output *linear* label functions $L_1(X), \dots, L_m(X)$ represented by their coefficient vectors $\mathbf{L} = (\mathbf{L}[1], \dots, \mathbf{L}[m])$
2. given **public** input \mathbf{x} , output label vector $\boldsymbol{\ell} = \mathbf{x} \cdot \mathbf{L} = (L_1(\mathbf{x}), \dots, L_m(\mathbf{x}))$

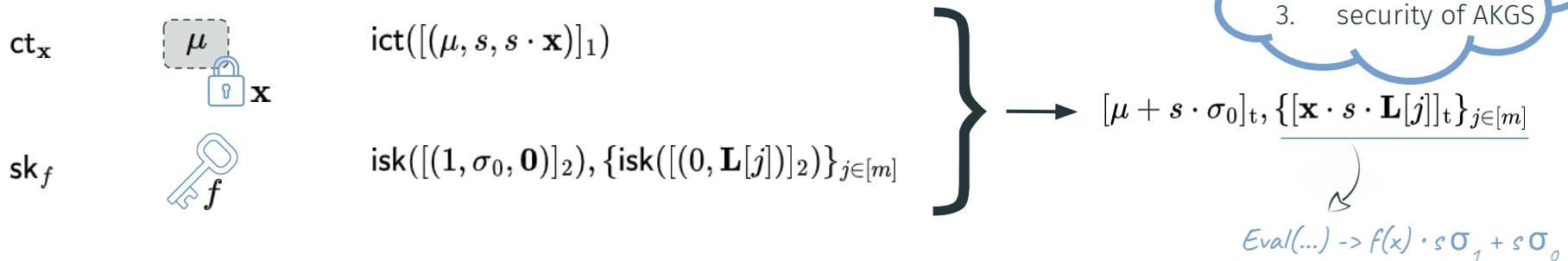
evaluation: given $f, \mathbf{x}, \boldsymbol{\ell}$, output $d = \sigma_1 f(\mathbf{x}) + \sigma_0$

simulation: given f, \mathbf{x}, d , output $\tilde{\boldsymbol{\ell}} \approx \boldsymbol{\ell}$

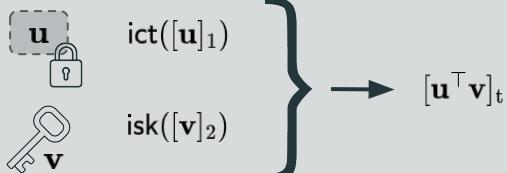
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Inner-Product Functional Encryption



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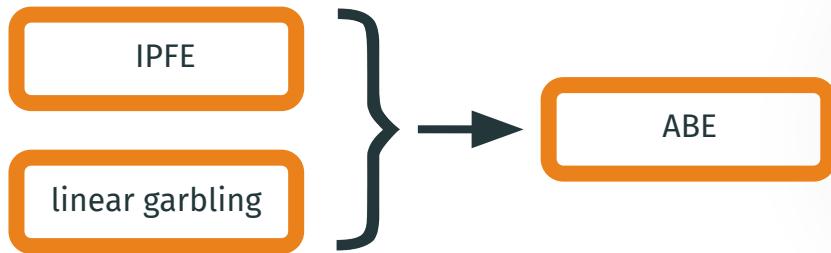
$\text{Eval}(f, \mathbf{x}, \ell := \mathbf{x} \cdot \mathbf{L}) \rightarrow d \text{ s.t. } d = \sigma_1 f(\mathbf{x}) + \sigma_0$

$\text{Sim}(f, \mathbf{x}, d) \rightarrow \tilde{\ell} \text{ s.t. } \tilde{\ell} \approx \ell$

Linearity:

- labels linear in (1) σ 's, \mathbf{r} , and (2) \mathbf{x}
- evaluation linear in labels

Framework for *Registered ABE*



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Challenges in the registered setting.

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 - aggregation is **deterministic**
 - encryption time is polylogarithmic in number of users (**compactness**)

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Linearity to the rescue.

$$\mathbf{L} = (\sigma_0, \sigma_1, \mathbf{r}) \cdot \widehat{\mathbf{L}}$$

-> **offline/online** phase

RFE for Pre-IP (Batch Variant)

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Theorem. Assuming bilateral MDDH on pairings, there exists a SIM-secure RFE scheme for Pre-IP.

Proof. (see the paper, based on IND-secure RFE for IP of [EC:ZLZ⁺24])

How to pick the matrices?

Arithmetic Key Garbling Scheme [EC:LL20]

$\text{Garble}(f, \sigma_0, \sigma_1; \mathbf{r}) \rightarrow \mathbf{L} = (\mathbf{L}[1], \dots, \mathbf{L}[m])$

$\text{Eval}(f, \mathbf{x}, \boldsymbol{\ell} := \mathbf{x} \cdot \mathbf{L}) \rightarrow d \text{ s.t. } d = \sigma_1 f(\mathbf{x}) + \sigma_0$

Labels are

- linear in secret input and randomness
- linear in public input

$$\boldsymbol{\ell} = (\mathbf{x} \otimes (\sigma_0, \sigma_1, \mathbf{r})) \cdot \widehat{\mathbf{L}}$$

How to pick the matrices?

$$[\mathbf{u}]_1 = [(\mu, s, \mathbf{x} \otimes s)]_1 \quad [\mathbf{P}_i]_2 = \left[\begin{pmatrix} 1 \\ \sigma_{i,0} \\ \mathbf{I}_{|\mathbf{x}|} \otimes (\sigma_{i,0}, \sigma_{i,1}, \mathbf{r}_i) \end{pmatrix} \right]_2 \quad \mathbf{V}_i = \begin{pmatrix} 1 \\ \widehat{\mathbf{L}}_i \end{pmatrix}$$

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Correctness. *RIPFE decryption yields* $[\mathbf{d}_i]_t = [\mathbf{u} \mathbf{P}_i \mathbf{V}_i]_t = \left[\left(\mu + s\sigma_{i,0}, (\mathbf{x} \otimes (s\sigma_{i,0}, s\sigma_{i,1}, s\mathbf{r}_i)) \cdot \widehat{\mathbf{L}}_i \right) \right]_t$

$\rightsquigarrow \text{Eval}(\dots) \rightarrow f_i(x) \cdot s\sigma_{i,1} + s\sigma_{i,0}$

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Security.

RIPFE leakage is $[\mathbf{d}_i]_t = [\mathbf{u} \mathbf{P}_i \mathbf{V}_i]_t = \left[\left(\mu + s\sigma_{i,0}, \underbrace{(\mathbf{x} \otimes (s\sigma_{i,0}, s\sigma_{i,1}, s\mathbf{r}_i)) \cdot \widehat{\mathbf{L}}_i}_{t,2} \right) \right]_t$

indistinguishable from $\text{Sim}(f, x, d \leftarrow \$)$

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$$\mathbf{V}_i = \begin{pmatrix} 1 \\ \widehat{\mathbf{L}}_i \end{pmatrix}$$

What about Turing machines?

Problem: shape of \mathbf{L} and \mathbf{r} depends on
input length, runtime and space
-> only known during encryption :/

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(input tape head, working tape head, working tape, state)

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$$\mathbf{T}(\mathbf{x})[c', c] = \begin{cases} 1 & \text{if } c \xrightarrow{M} c' \\ 0 & \text{otherwise} \end{cases}$$

c' = (k', j', w', q') *c = (k, j, w, q)*

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- we have $\mathbf{T}(\mathbf{x}) \cdot \mathbf{e}_c^\top = \mathbf{e}_{c'}^\top$

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- we have $\mathbf{T}(\mathbf{x}) \cdot \mathbf{e}_c^\top = \mathbf{e}_{c'}^\top$ and more general

$$M|_{N,S,T}(\mathbf{x}) = (\mathbf{1} \otimes \mathbf{y}_{\text{acc}}) \cdot (\mathbf{T}(\mathbf{x}))^T \cdot \mathbf{e}_{c_0}^\top$$

Arithmetization of TM Computations

- consider TM $M = (Q, \mathbf{y}_{\text{acc}}, \delta)$ and denote by (N, S, T) the input length, space and runtime
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... so we only need to garble matrix multiplication

Arithmetic Key Garbling for Logspace TMs [EC:LL20]

- **garbling:** sample $\mathbf{r}_0, \mathbf{r}_1, \dots, \mathbf{r}_T \leftarrow_{\$} \mathbb{Z}_p^c$ and output label functions

$$L_{\text{init}}(\mathbf{x}) = \sigma_0 + \mathbf{r}_0 \cdot \mathbf{e}_{c_0}^\top$$

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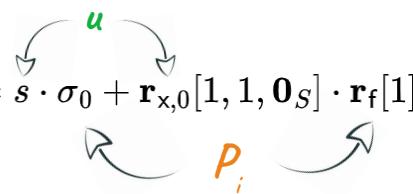
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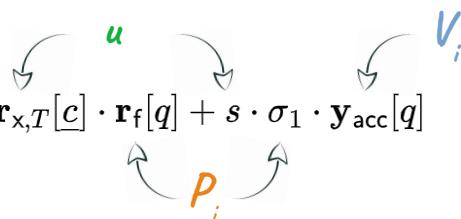
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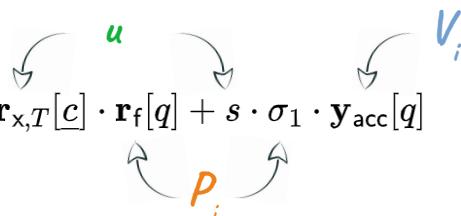
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✓

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✗

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Block Structure of Transition Matrix

- transition matrix $\mathbf{T}(\mathbf{x})[c', c] = \begin{cases} 1 & \text{if } \delta(q, \mathbf{x}[k], \mathbf{w}[j]) = (q', \mathbf{w}'[j], k' - k, j' - j), \mathbf{w}[\neq j] = \mathbf{w}'[\neq j] \\ 0 & \text{otherwise} \end{cases}$

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- consider $Q \times Q$ blocks $\mathbf{T}(\mathbf{x})[\underbrace{(k', j', \mathbf{w}', _)}, \underbrace{(k, j, \mathbf{w}, _)}]$

$$\mathbf{T}(\mathbf{x}) = \left(\begin{array}{c|c} & \text{block column } (c = (k, j, \mathbf{w}), _) \\ \hline \text{block row } (c' = (k', j', \mathbf{w}'), _) & \end{array} \right)$$

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\mathbf{u} \mathbf{v}_i
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Generalization to RFE

- so far, we used σ_0 as a pad for (a fixed message) μ and σ_1 as a masking term

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$\text{Garble}(f, \sigma_0, \sigma_1; \mathbf{r}) \rightarrow \mathbf{L} = (\mathbf{L}[1], \dots, \mathbf{L}[m])$

$\text{Eval}(f, \mathbf{x}, \ell := (1, \mathbf{x}) \cdot \mathbf{L}) \rightarrow d \text{ s.t. } d = \sigma_1 f(\mathbf{x}) + \sigma_0$

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- more general, we can
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 - use σ_0 as pad for any other (independently computed) RFE functionality
-> attribute-based functionalities (AB-AWS, AB-QF)

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Thank you!!! :)