

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

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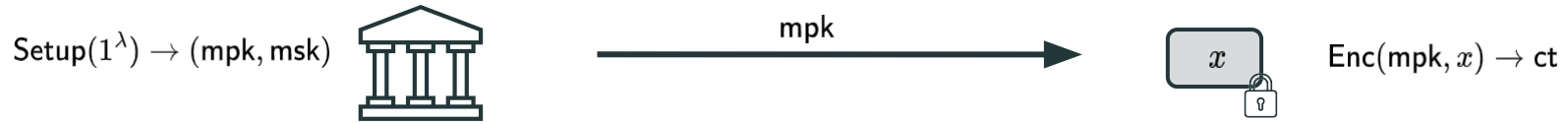


# Functional Encryption (FE) [TCC:BSW11]

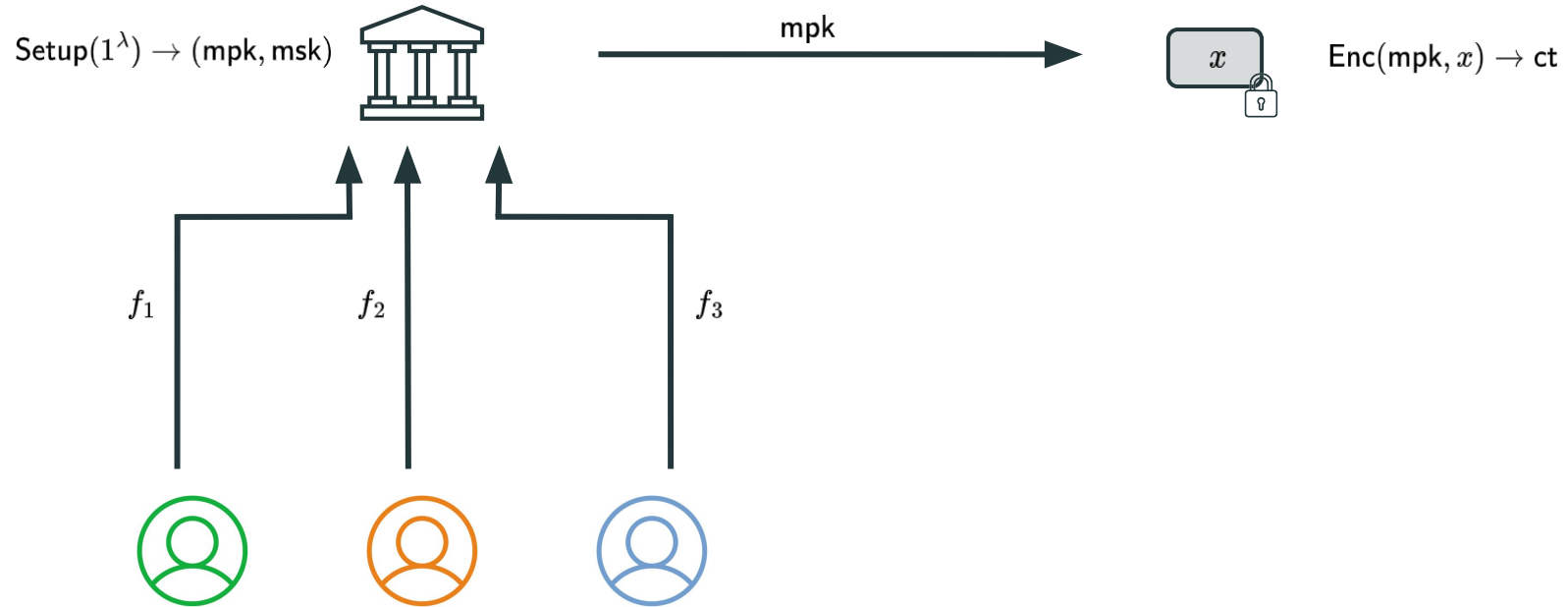
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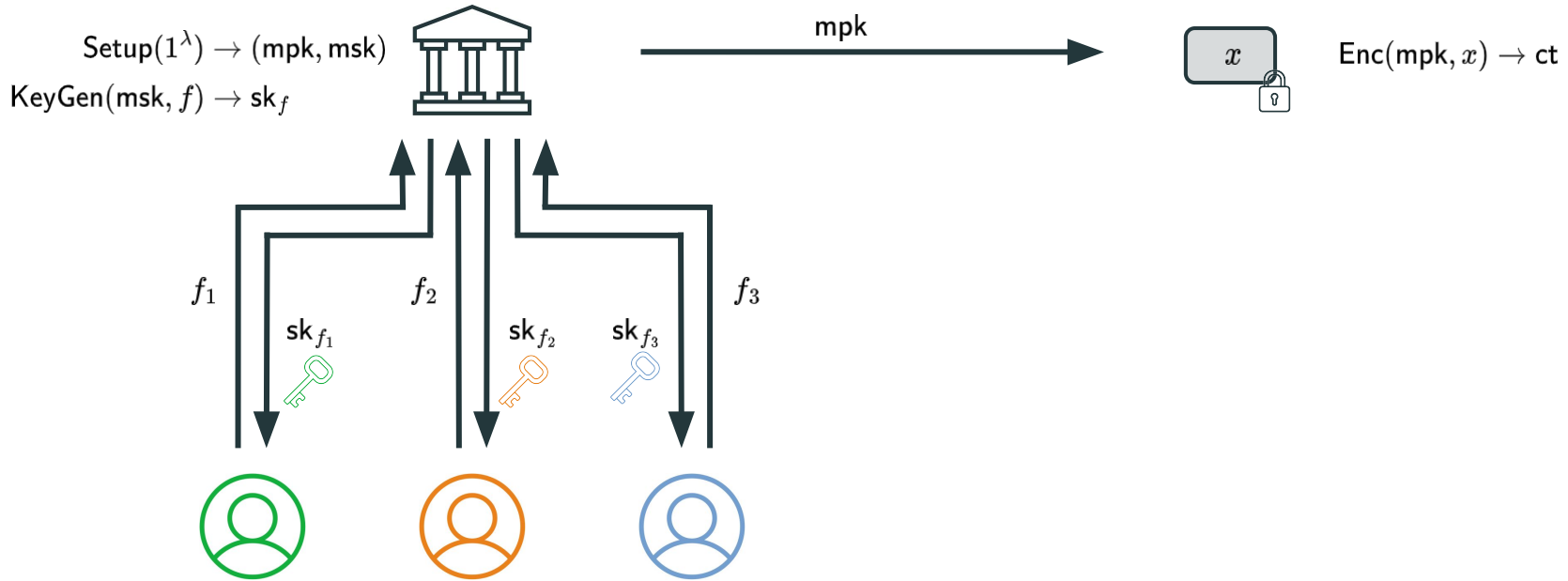
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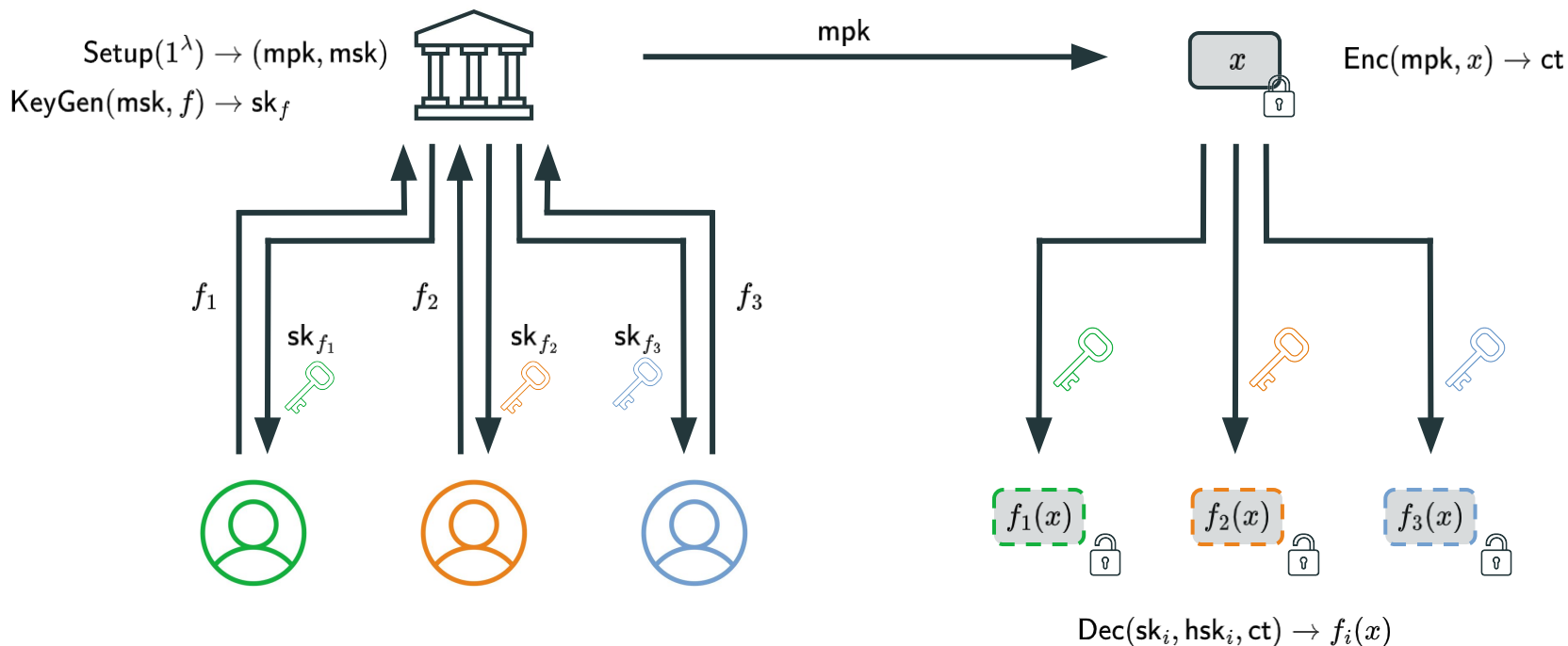
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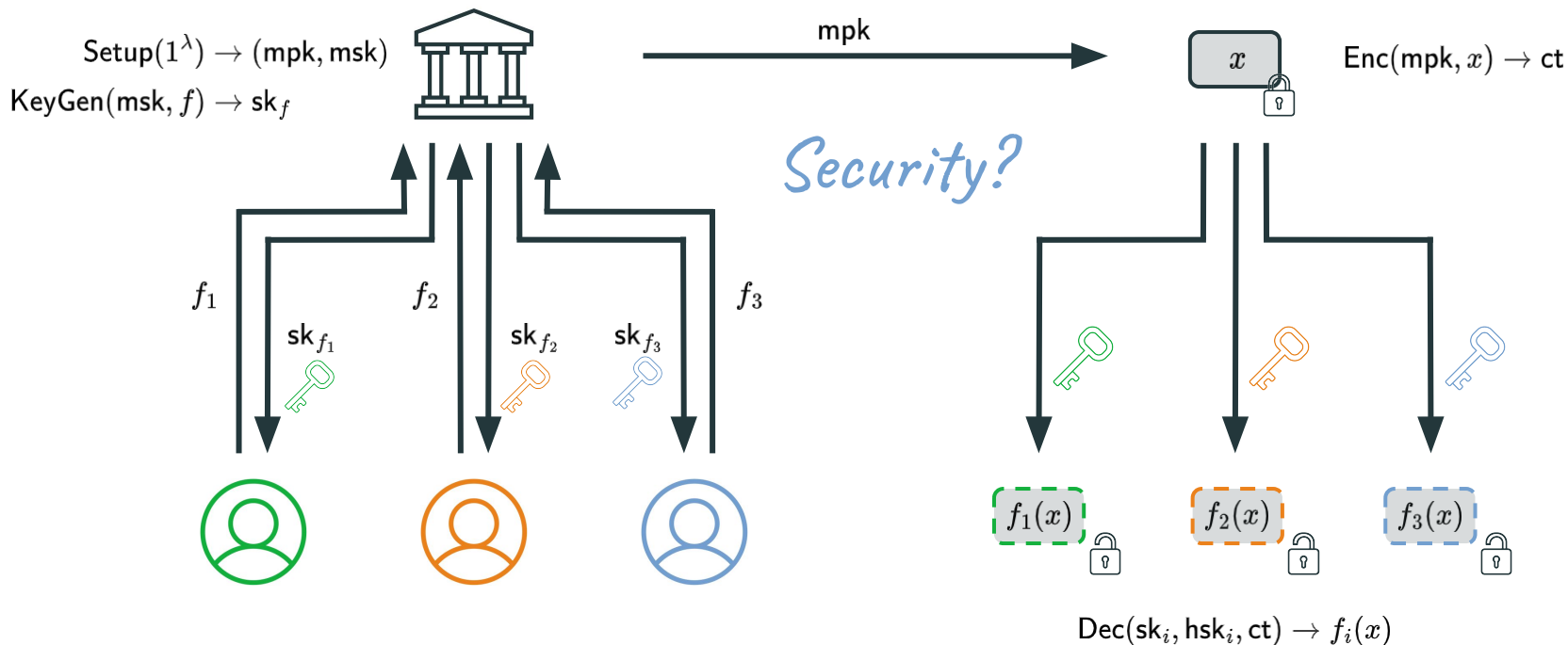
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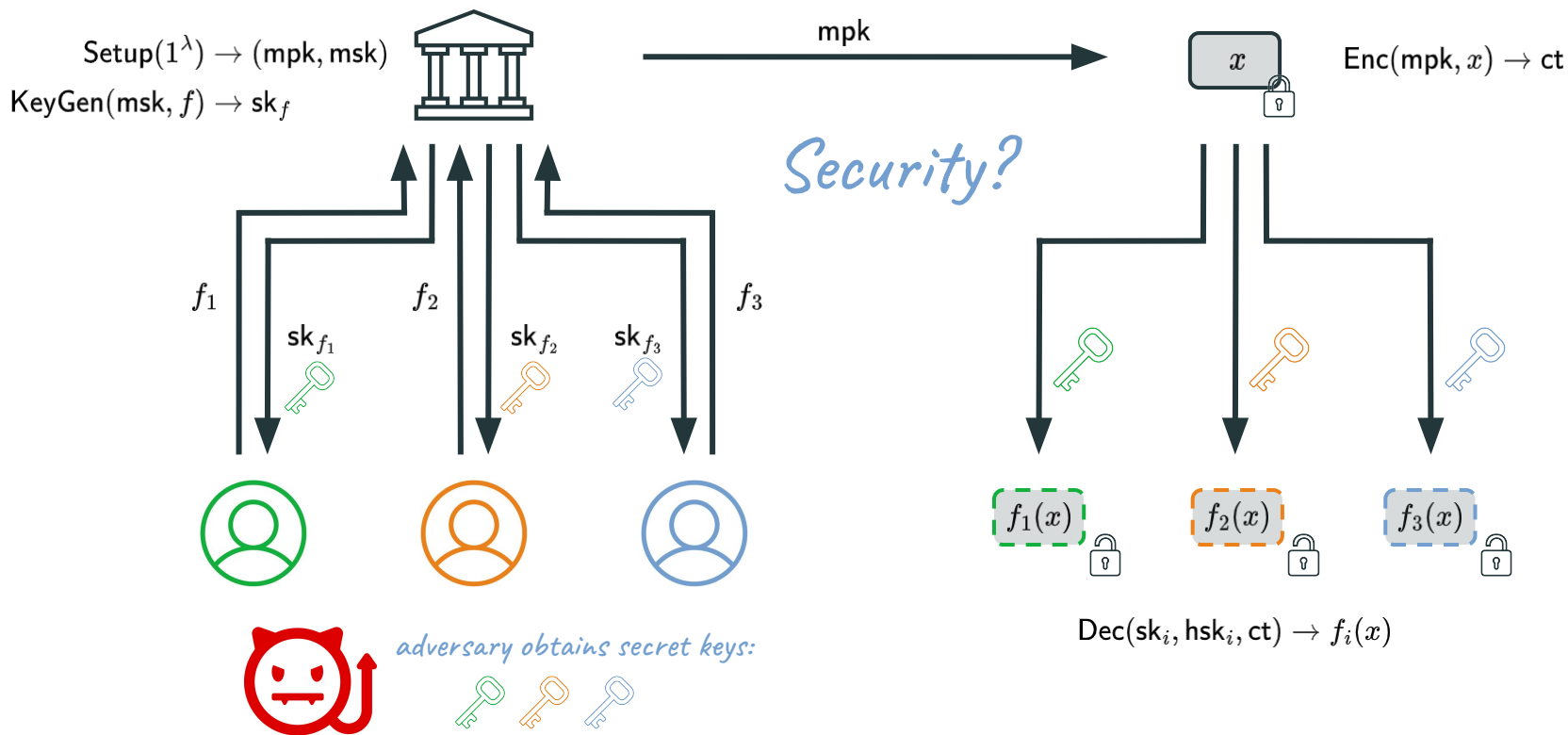
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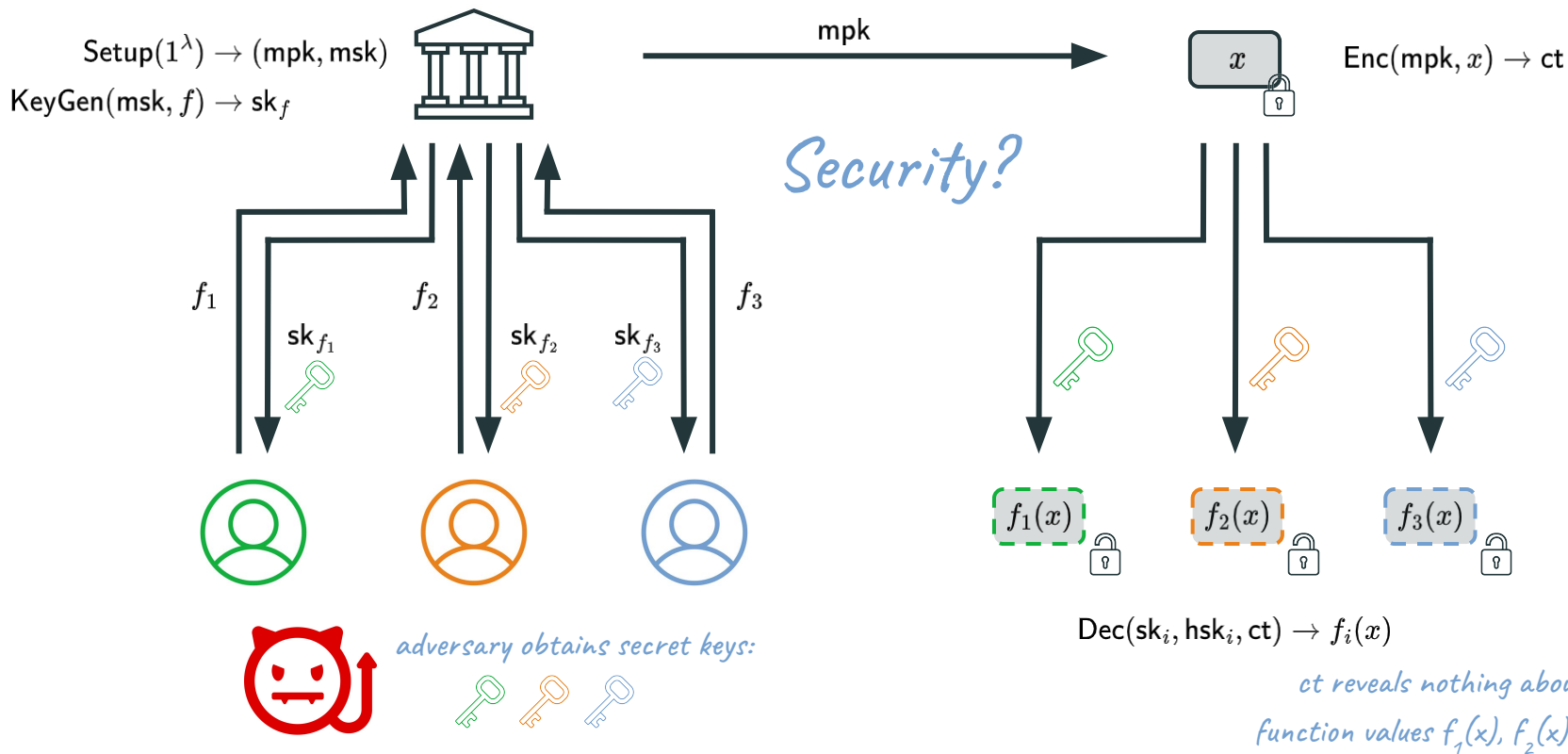


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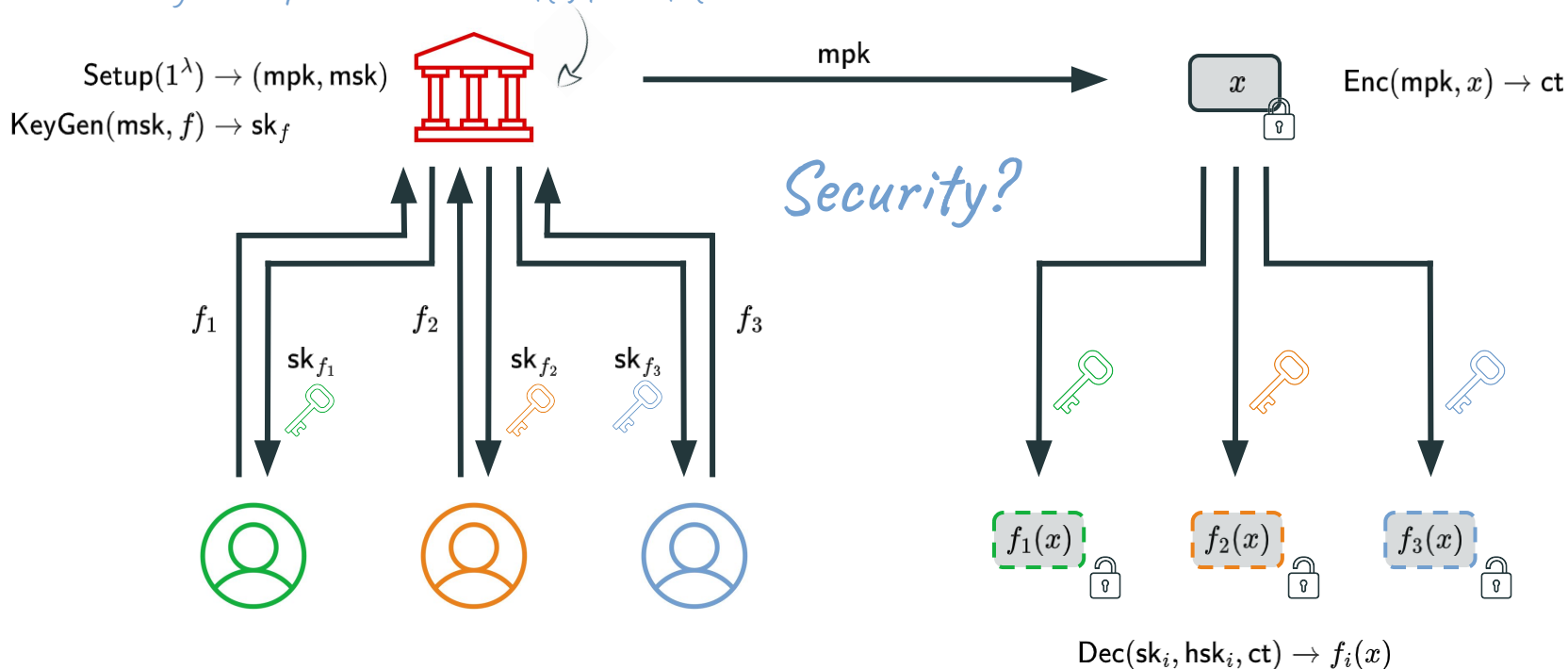


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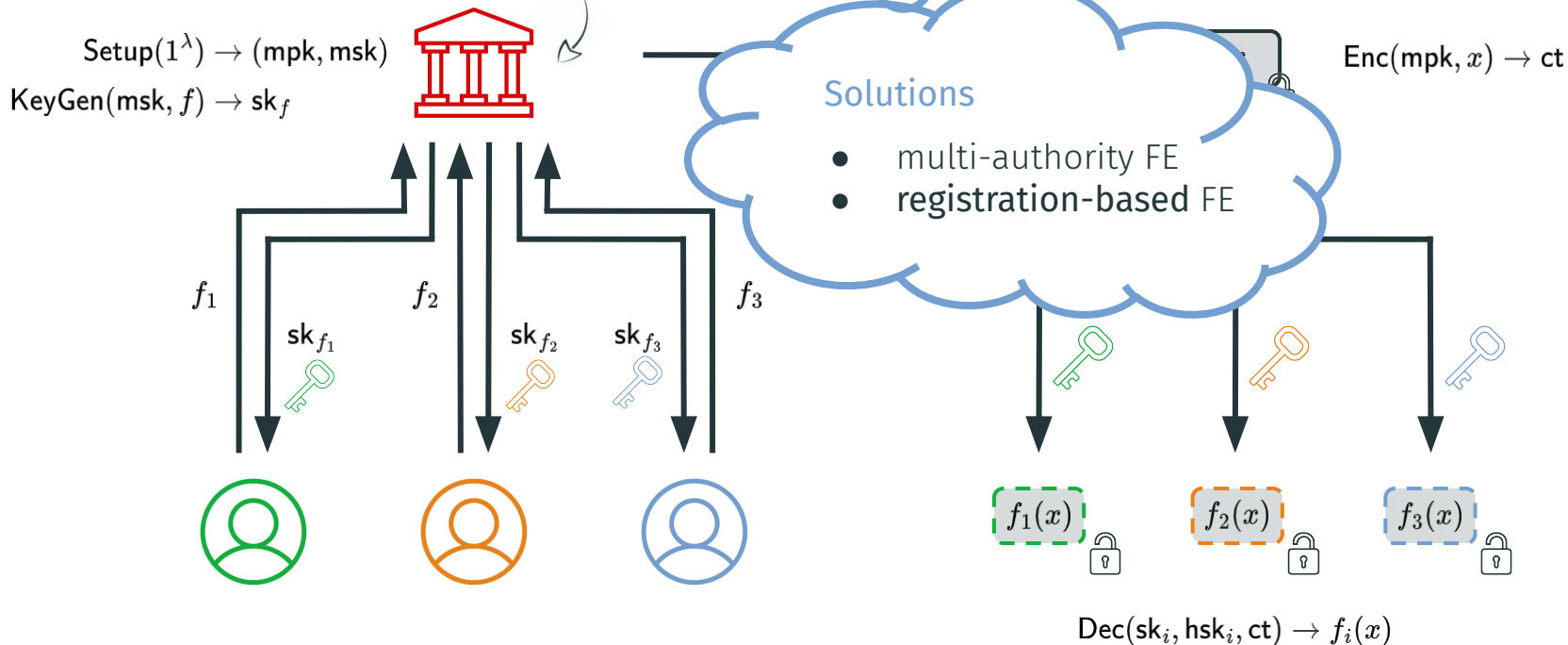
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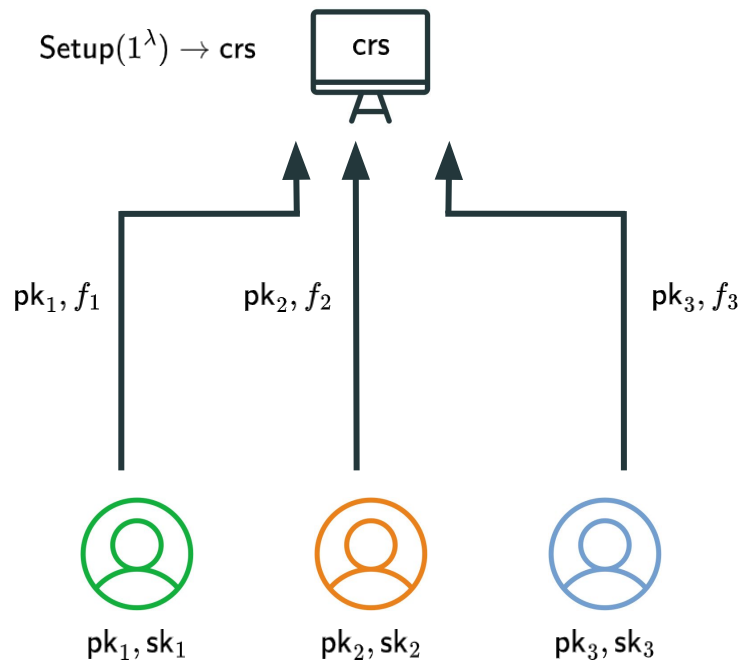
$\text{pk}_2, \text{sk}_2$



$\text{pk}_3, \text{sk}_3$

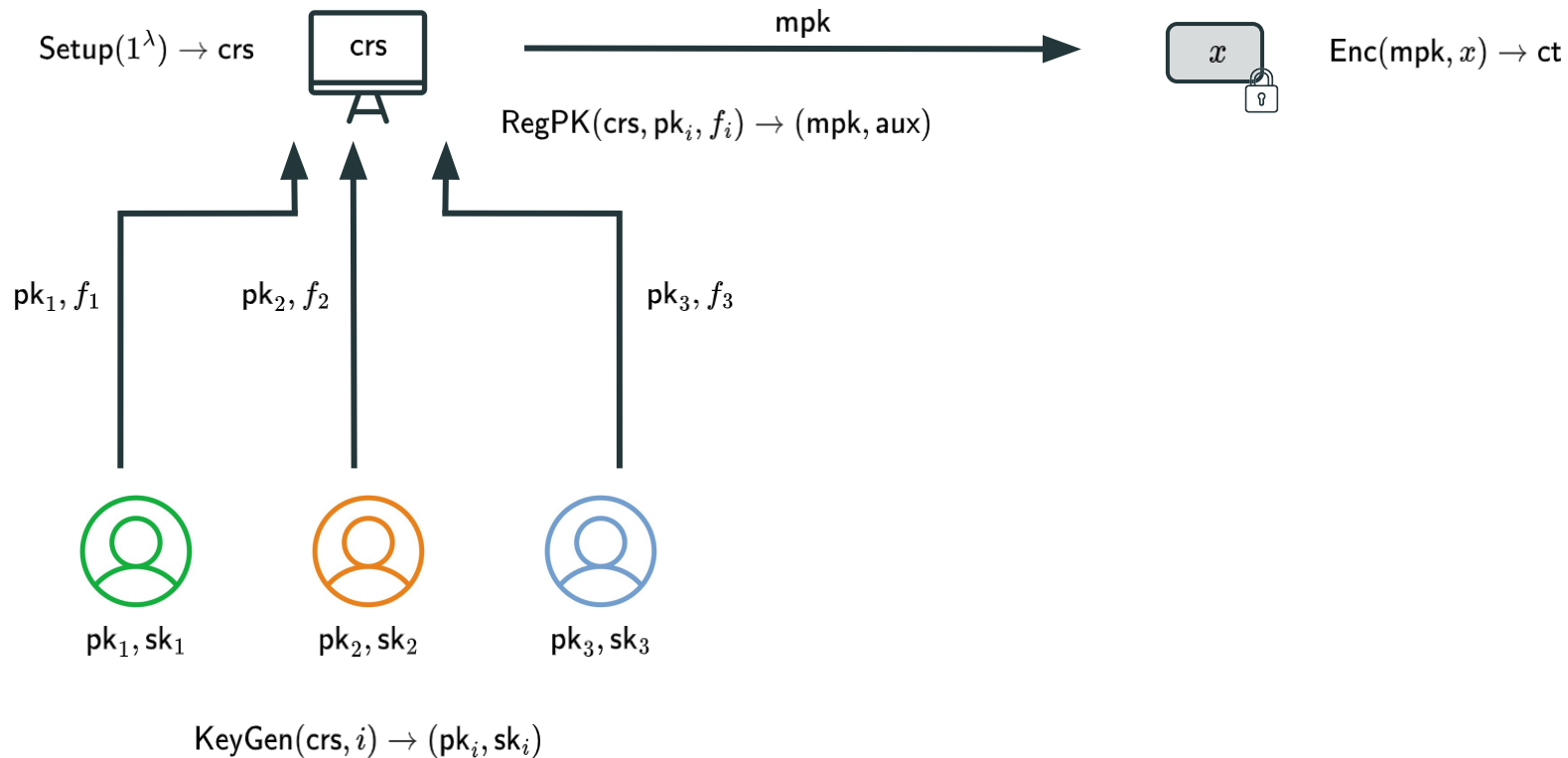
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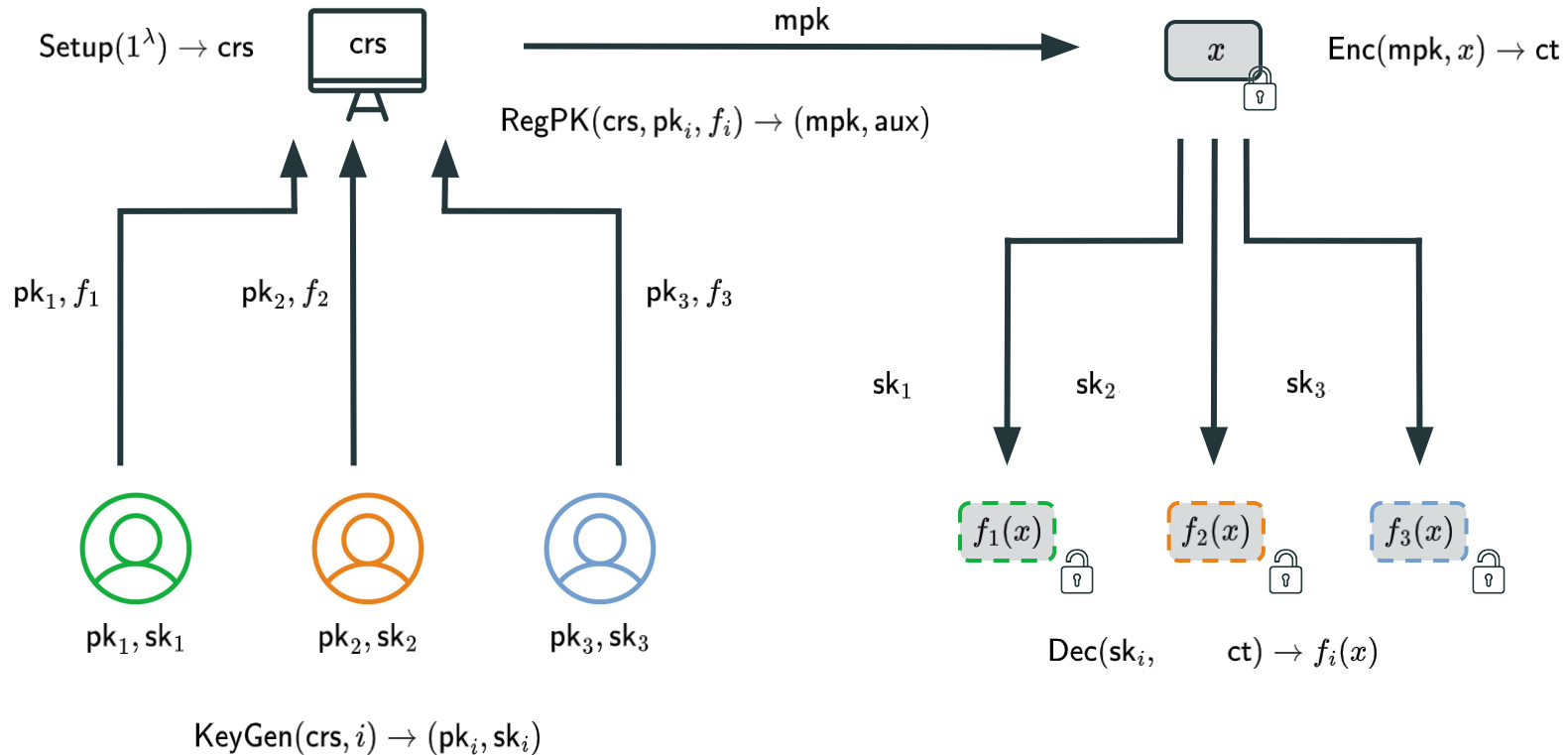


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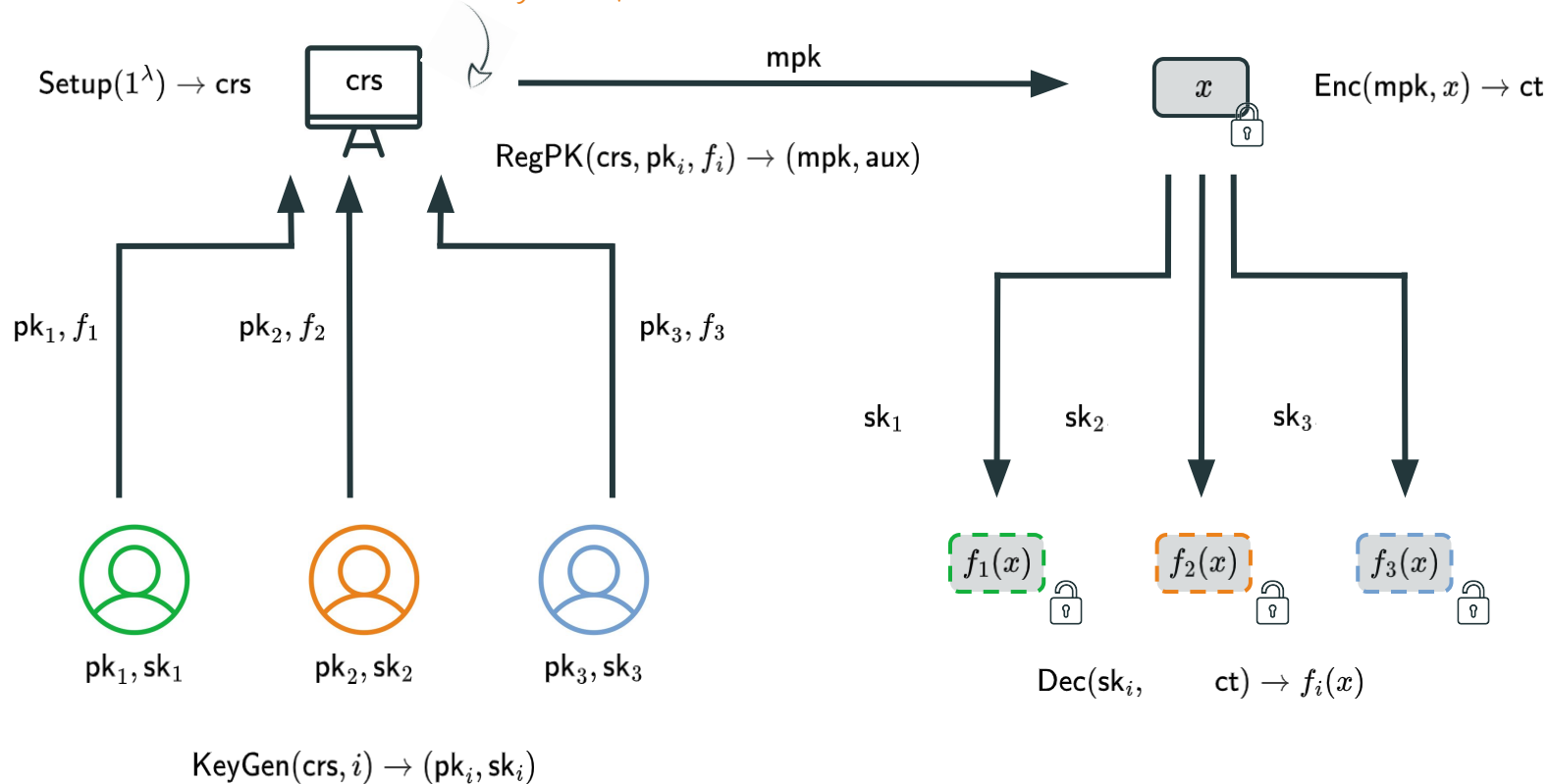


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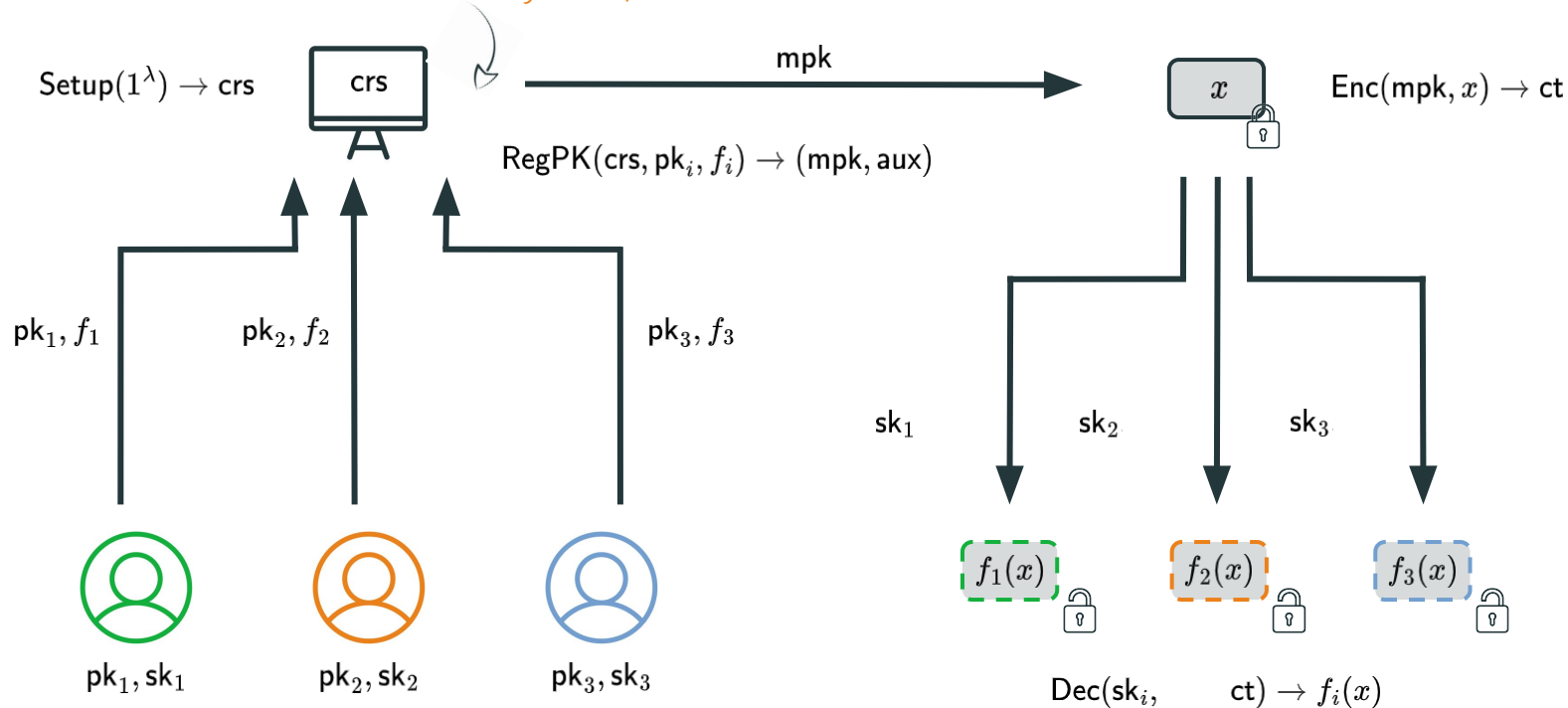
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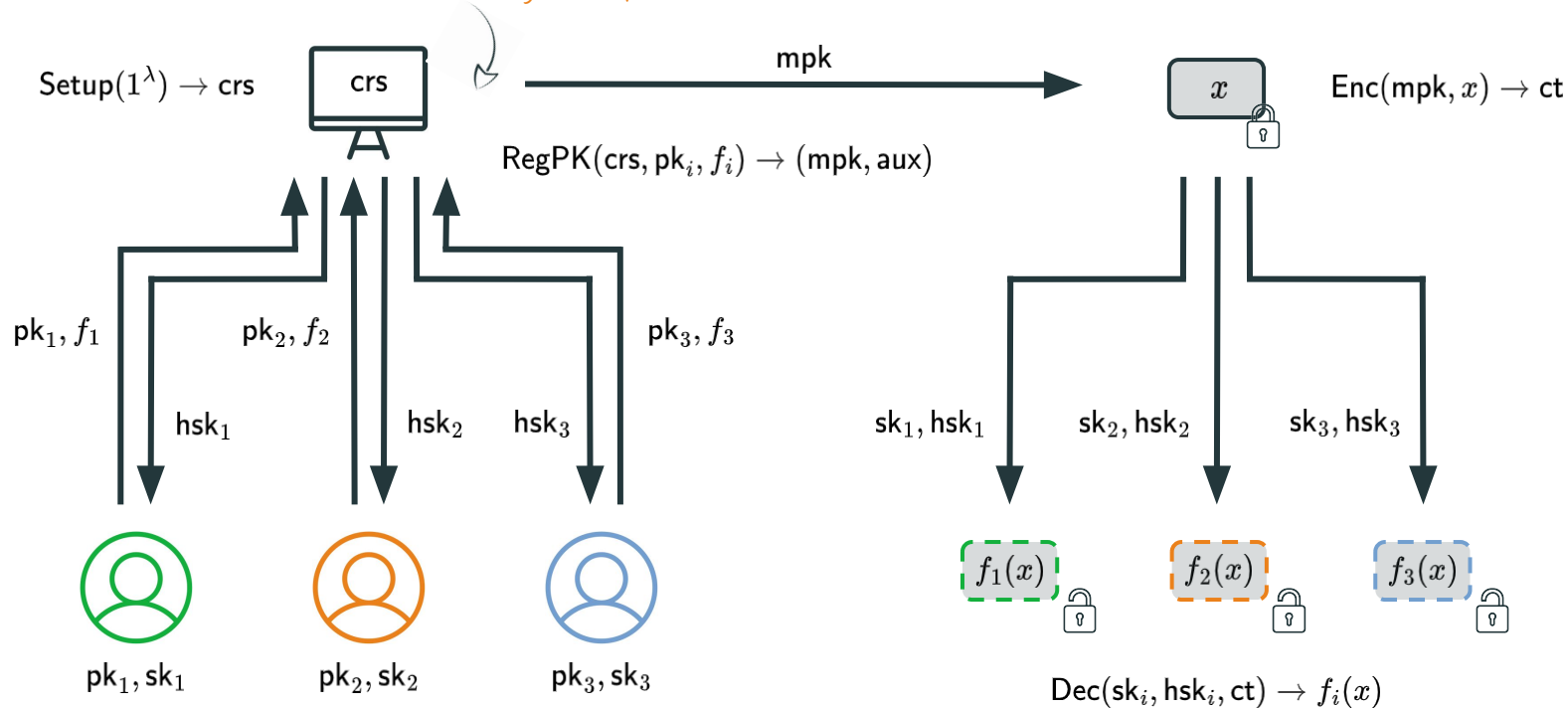


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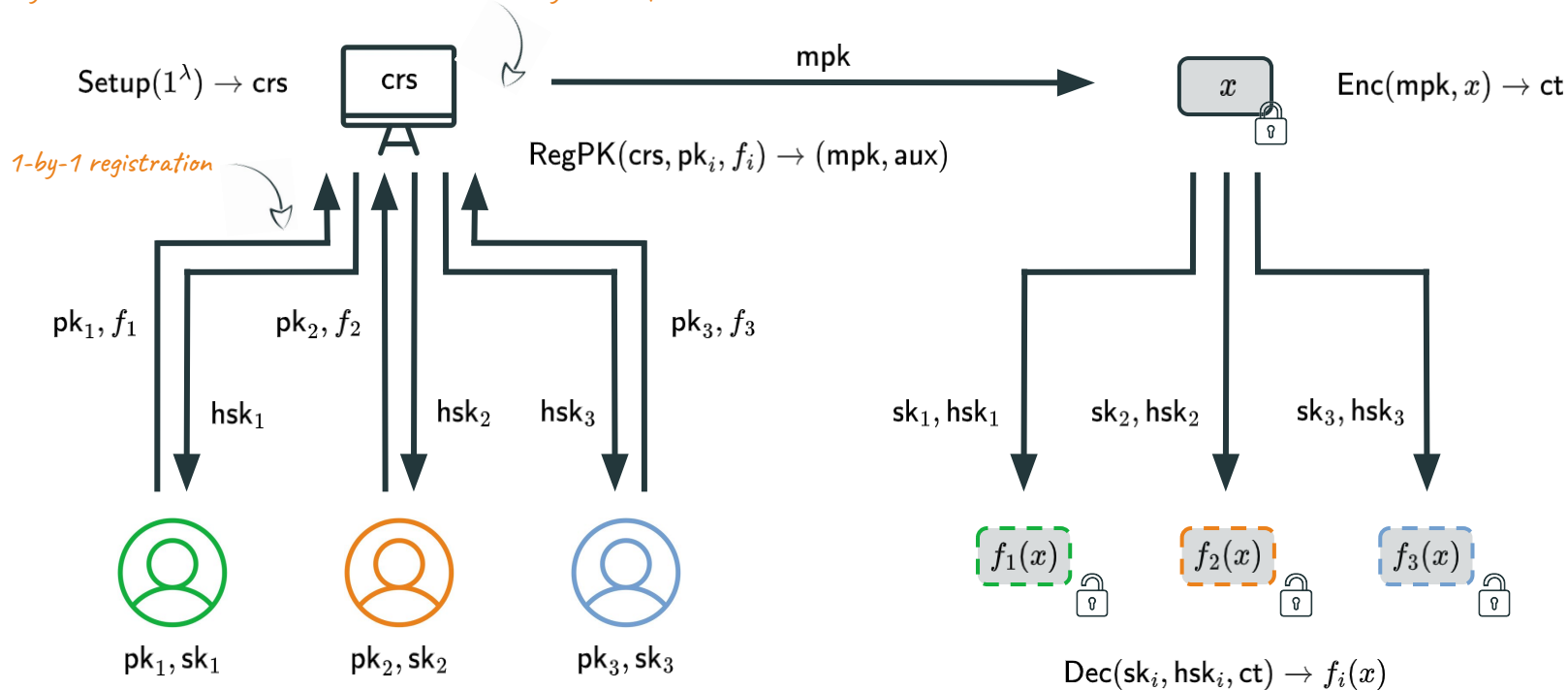
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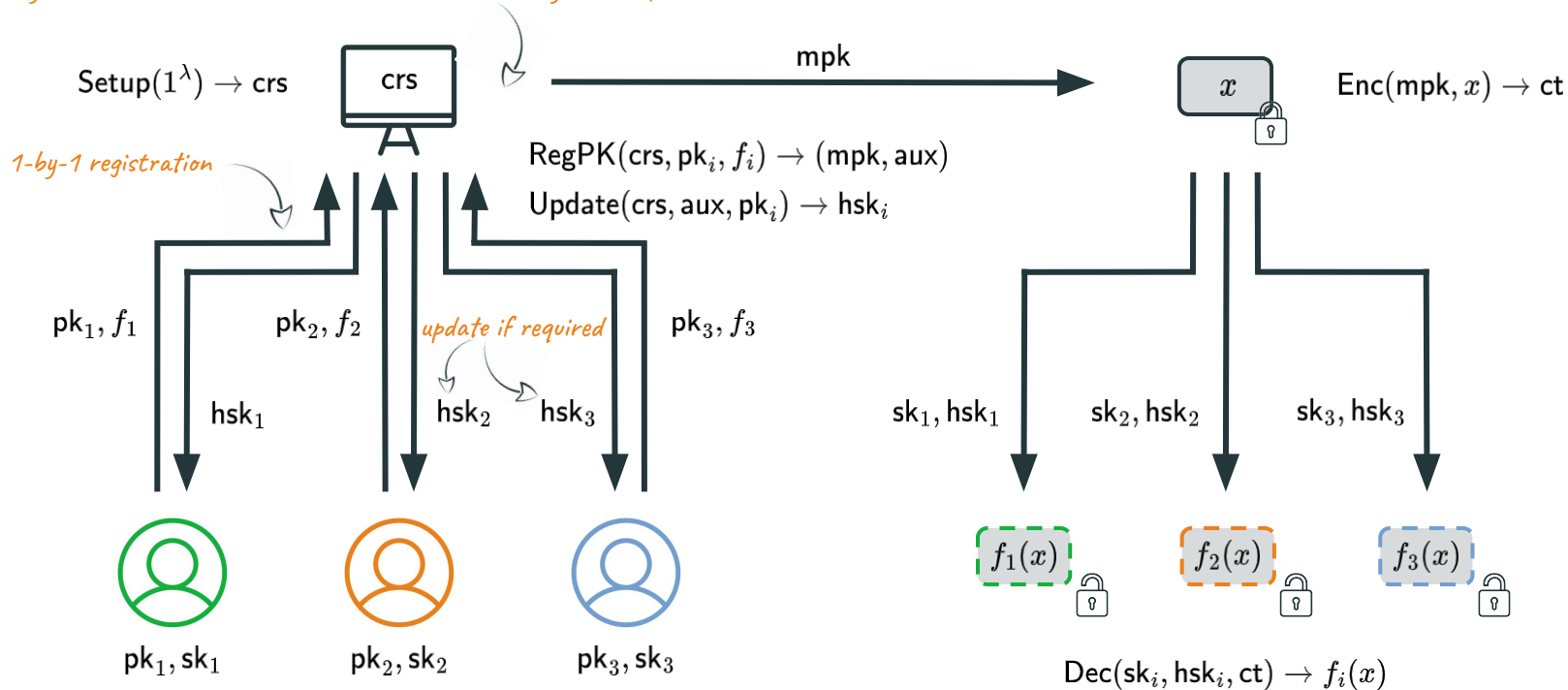


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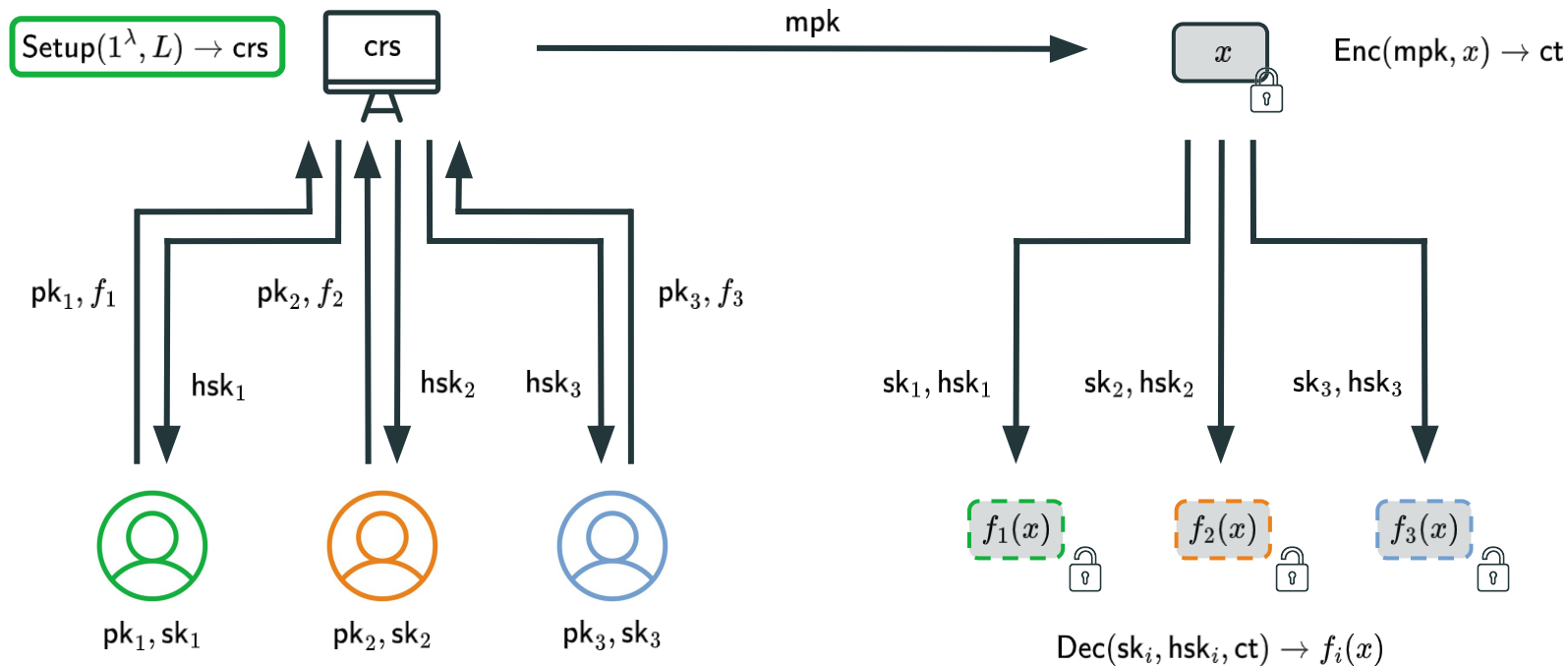
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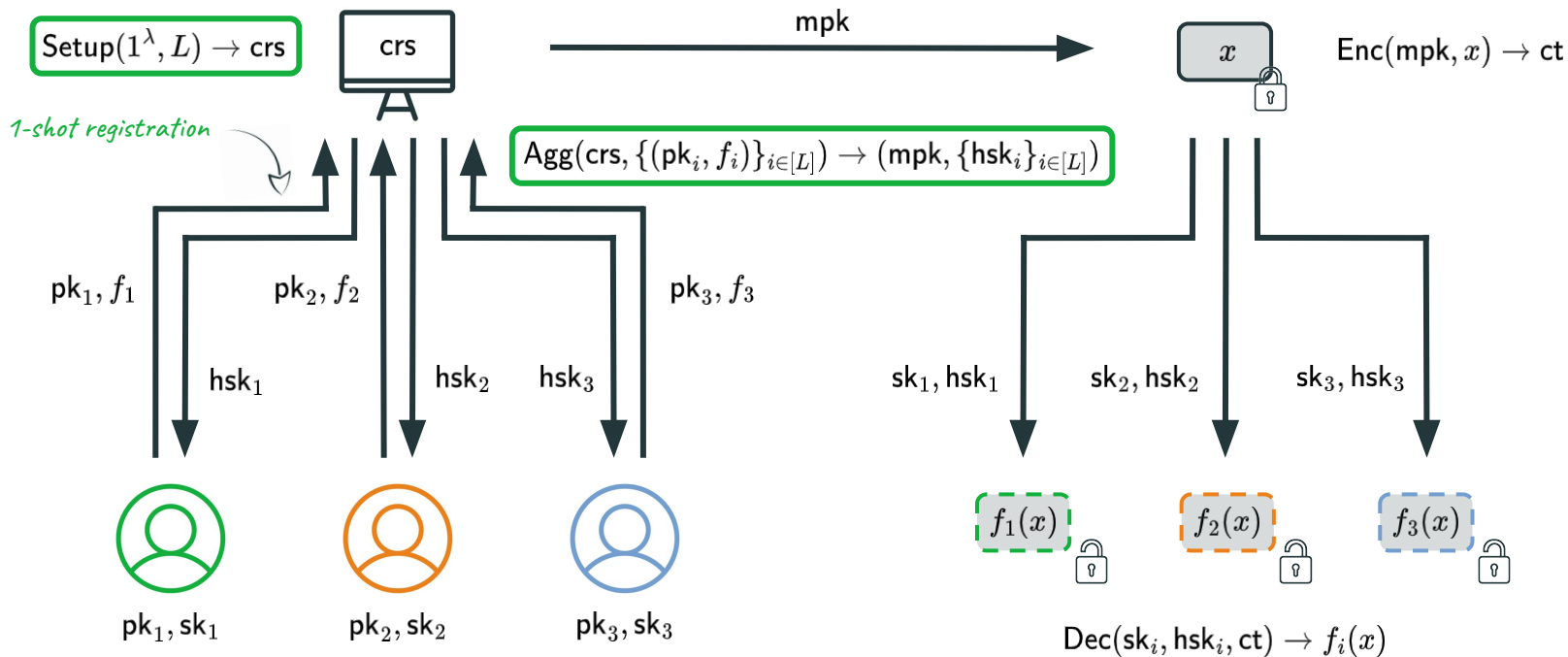
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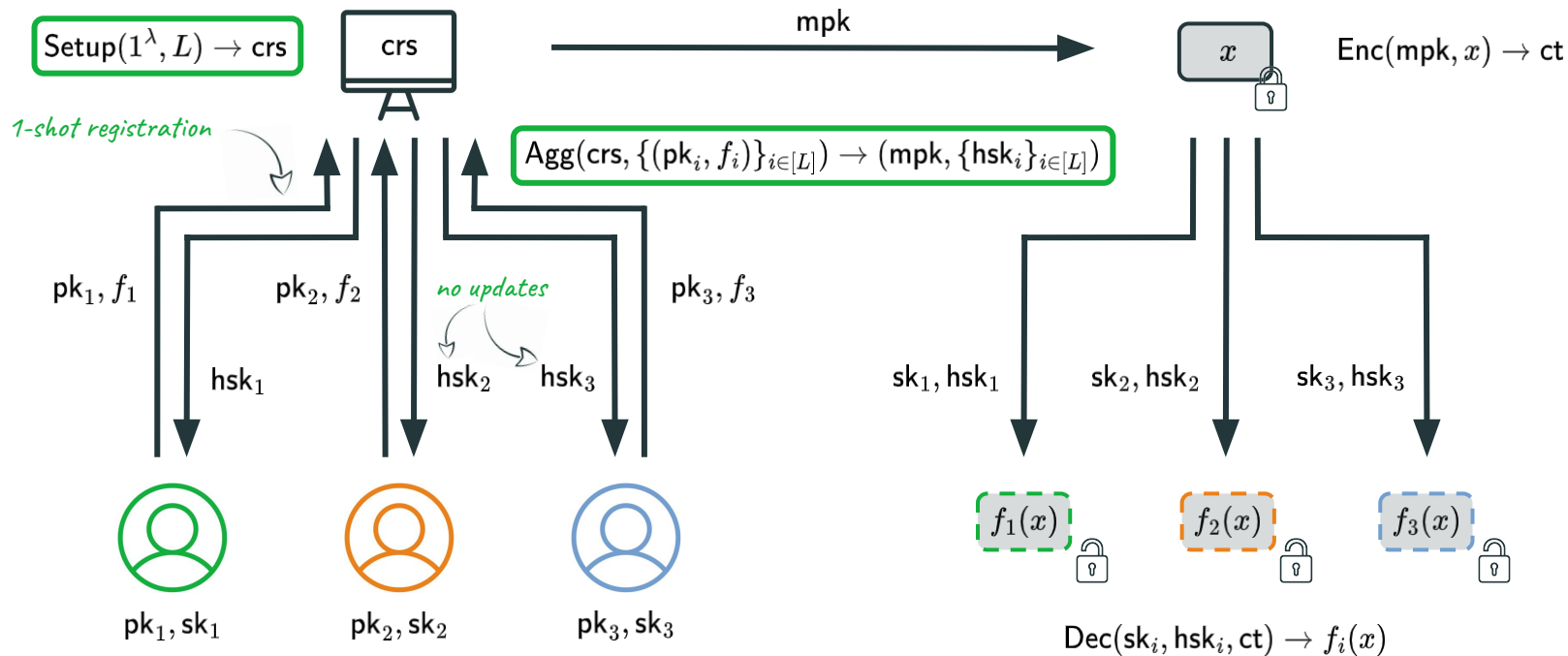
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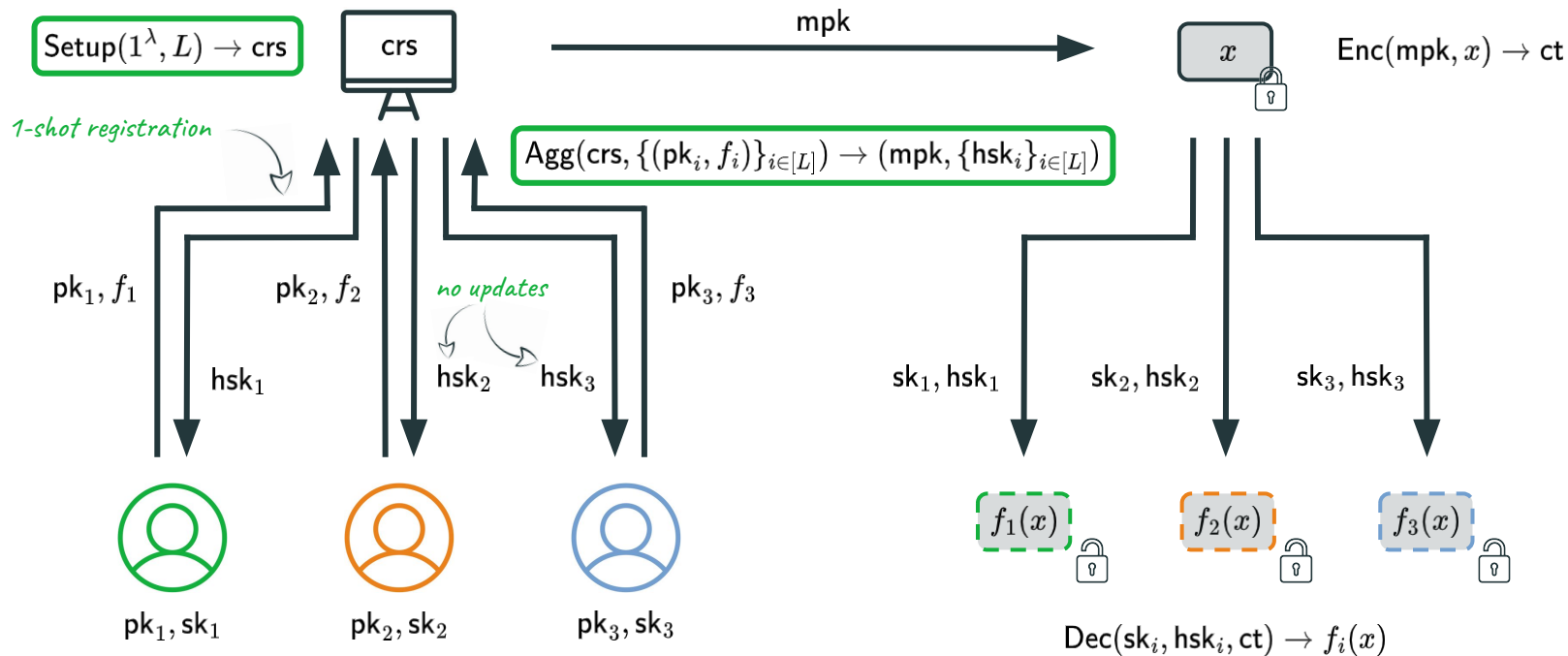


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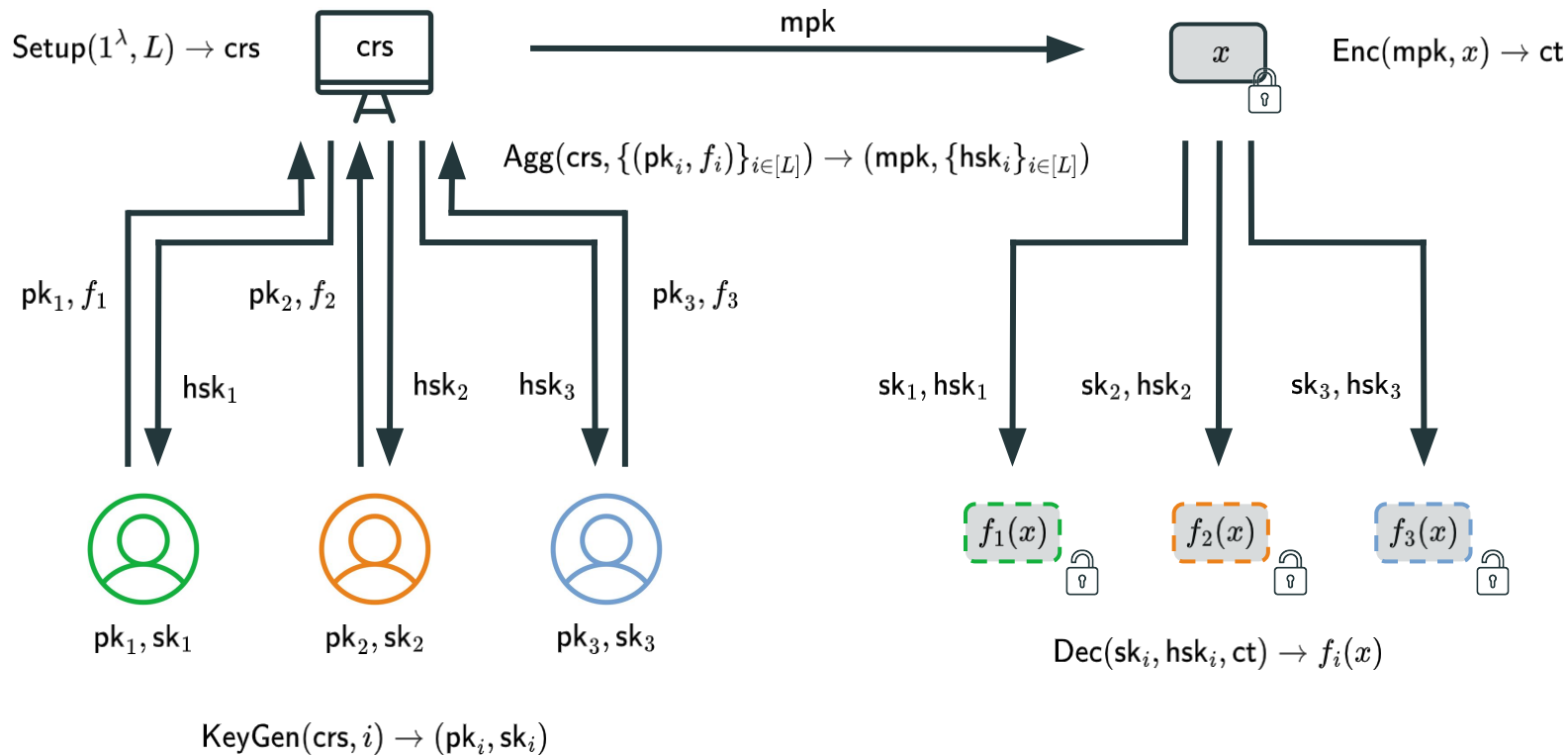
[HLWW23]: sRFE  $\Rightarrow$  RFE ("powers-of-two compiler")



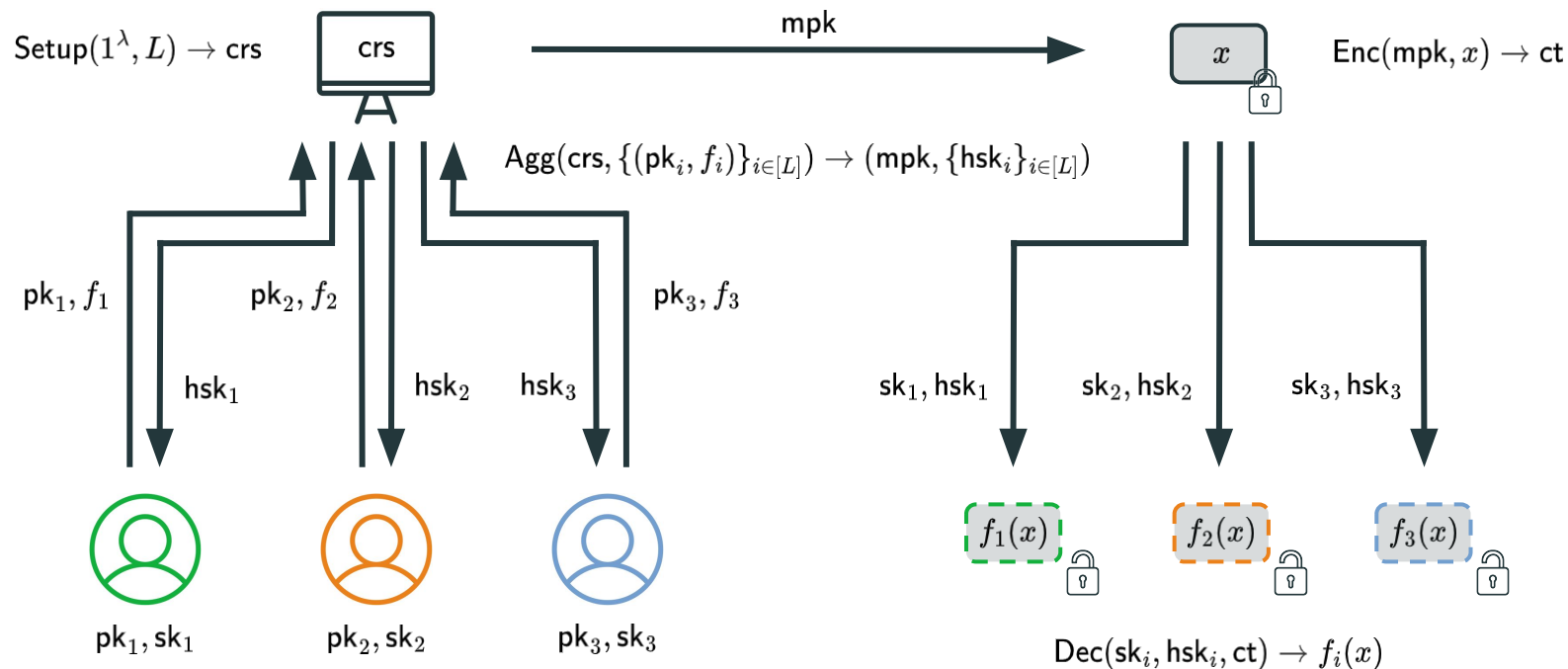
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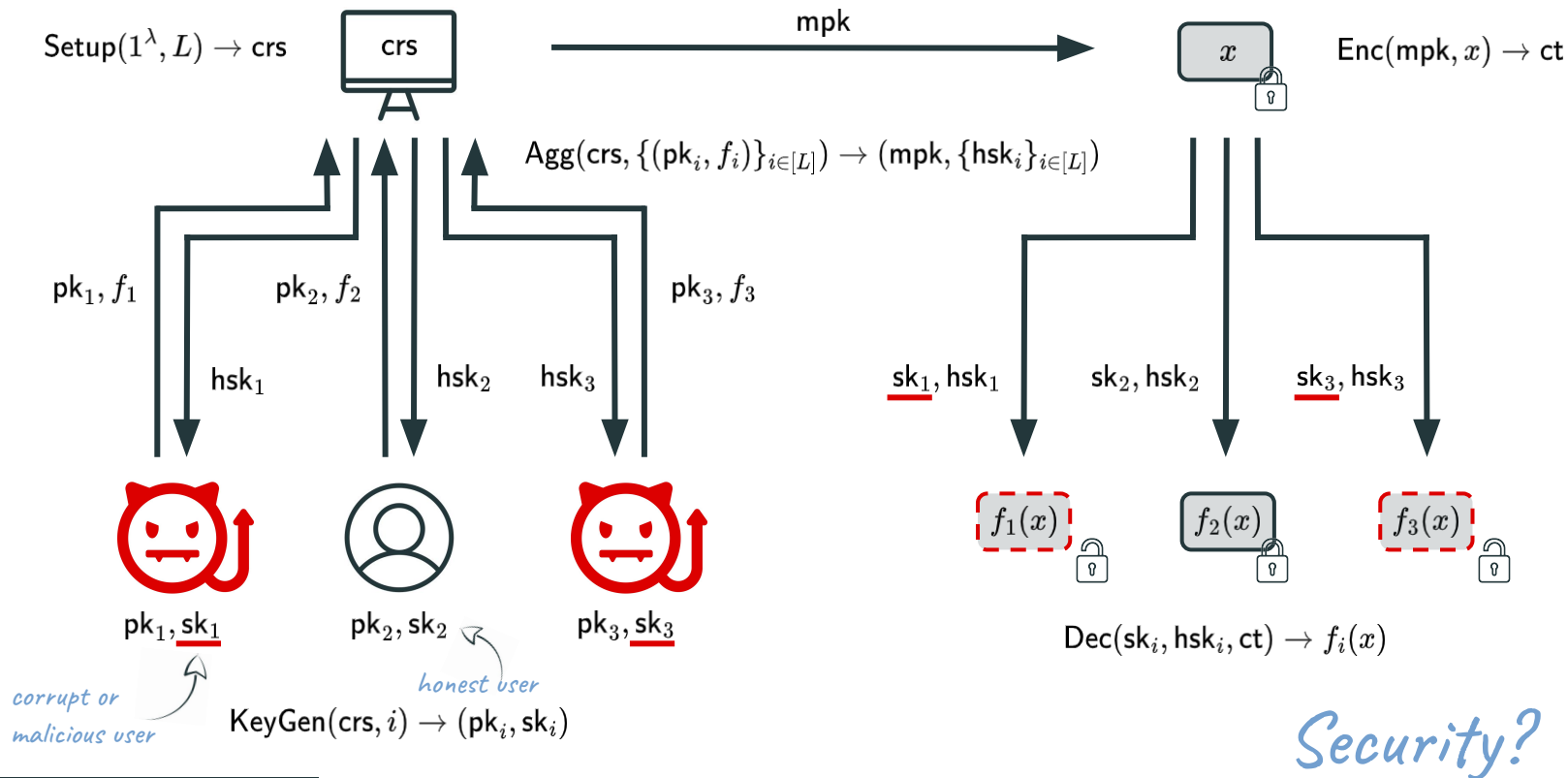
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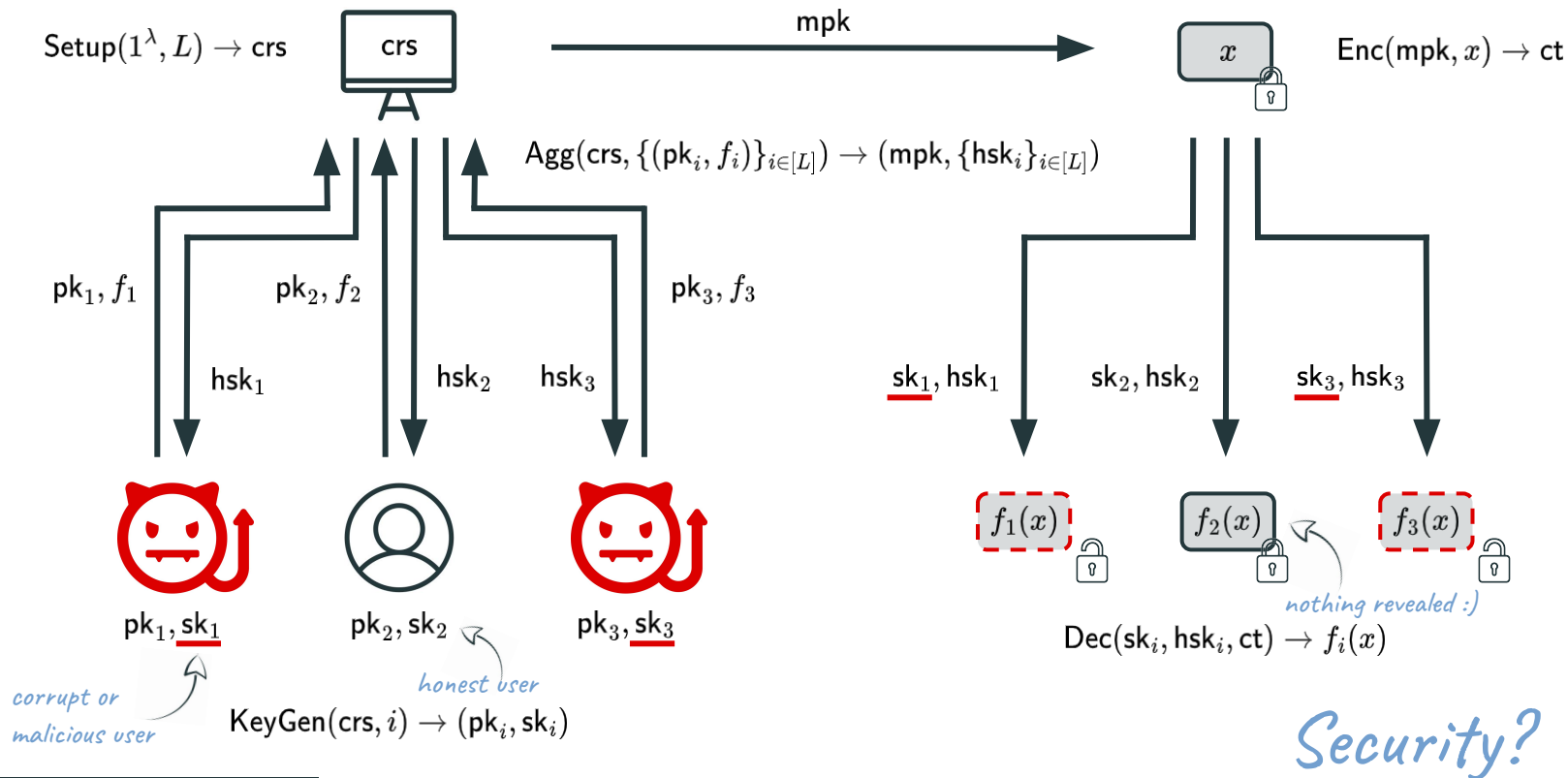
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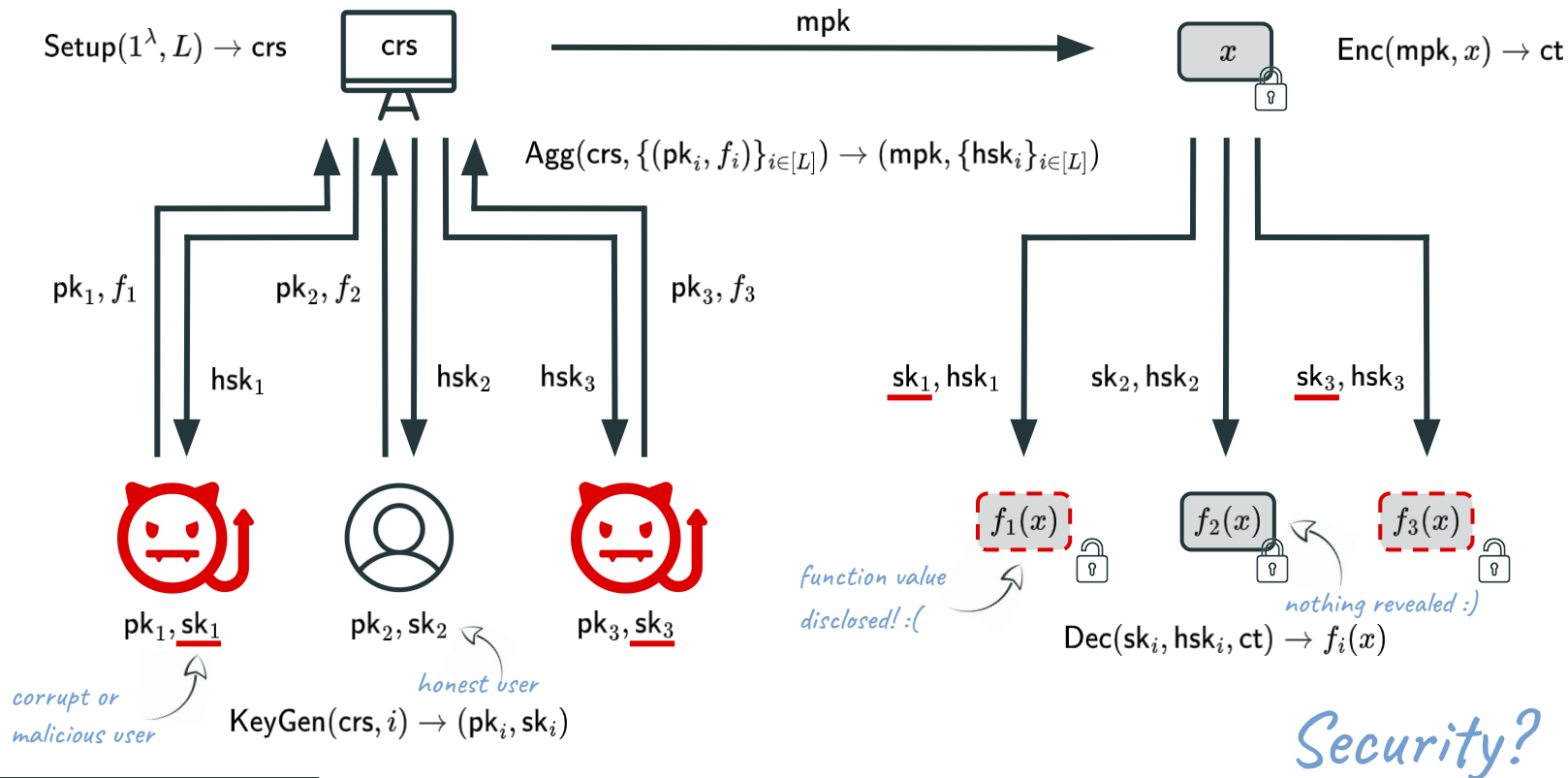
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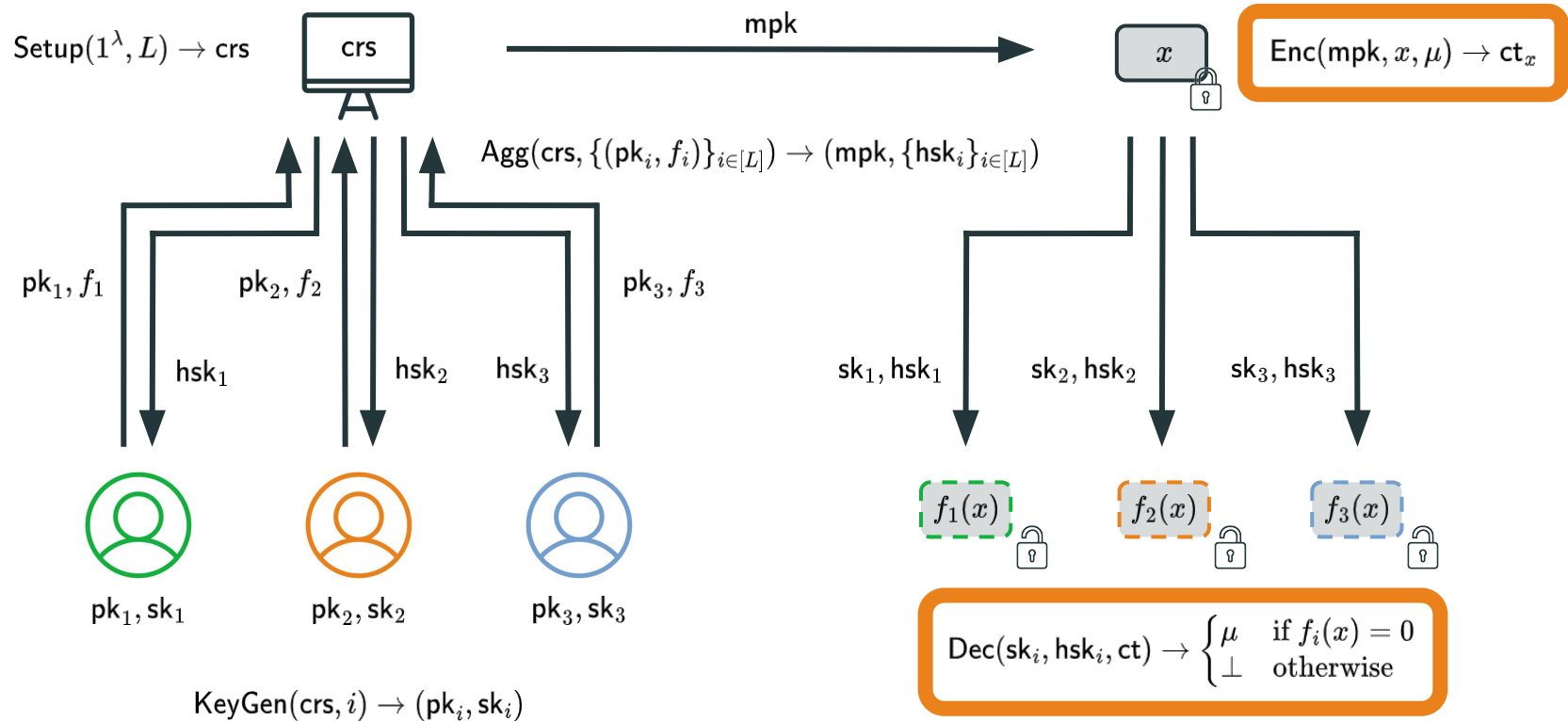
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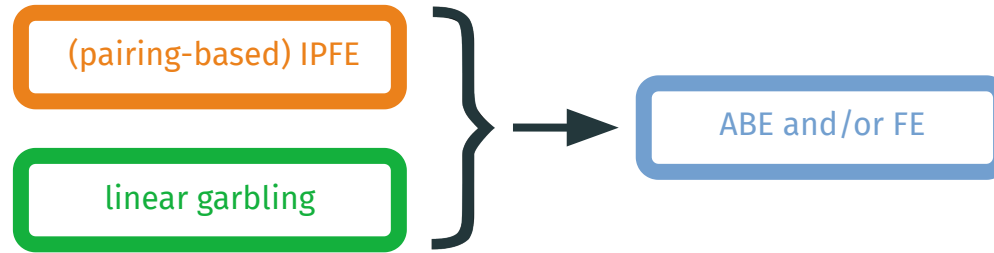
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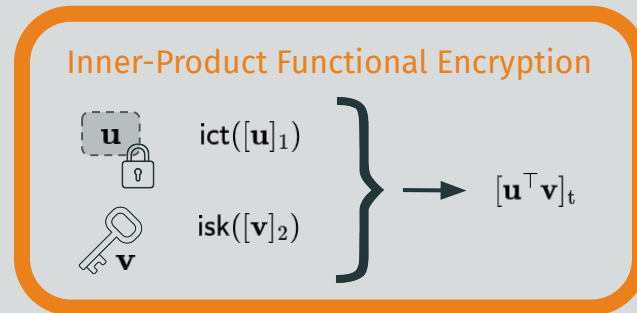
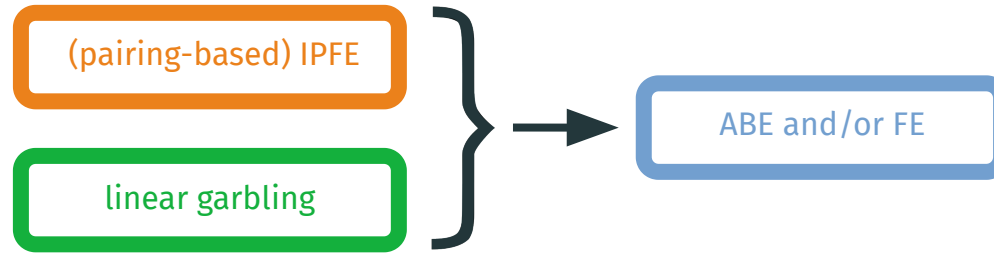
# Special Case: Registered ABE



# Framework for *Non-Registered* ABE

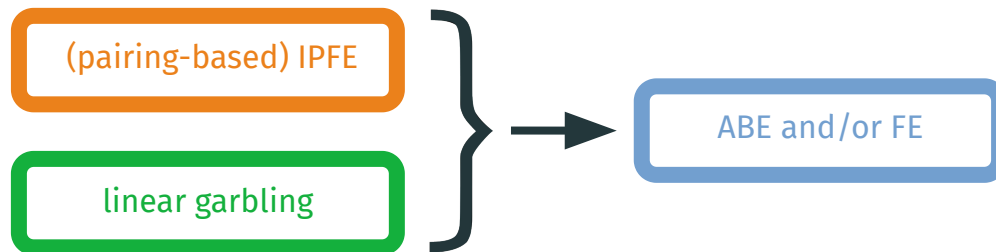


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## Arithmetic Key Garbling Scheme [EC:LL20]

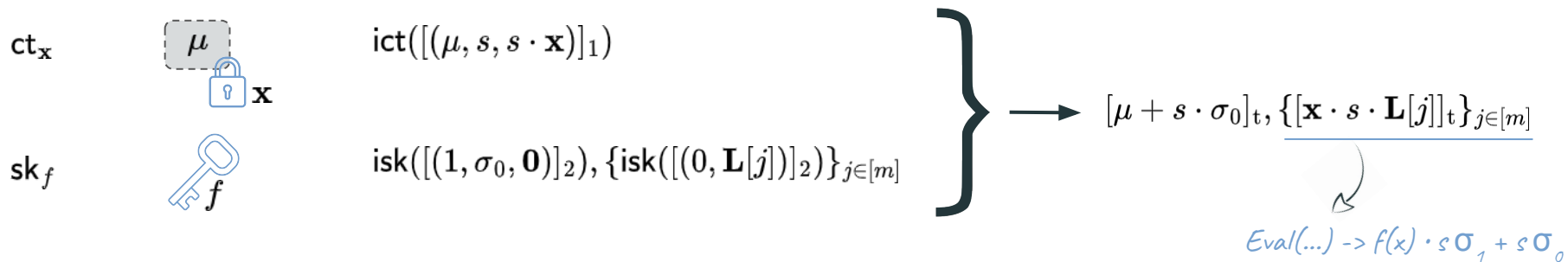
2-step garbling procedure:

1. given  $f$  and **secret** inputs  $\sigma_0, \sigma_1$ , 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  $\boldsymbol{\ell} = \mathbf{x} \cdot \mathbf{L} = (L_1(\mathbf{x}), \dots, L_m(\mathbf{x}))$

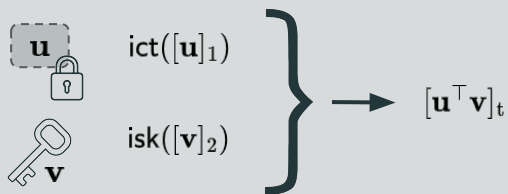
evaluation: given  $f, \mathbf{x}, \boldsymbol{\ell}$ , output  $d = \sigma_1 f(\mathbf{x}) + \sigma_0$

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

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## Inner-Product Functional Encryption



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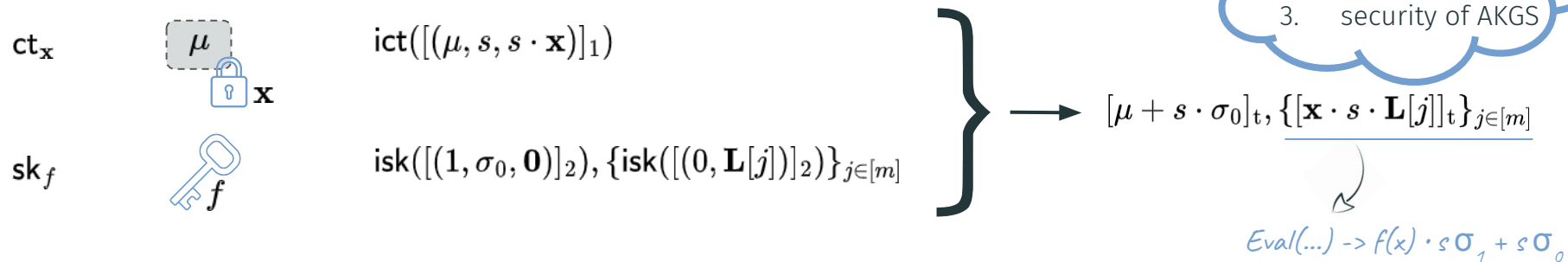
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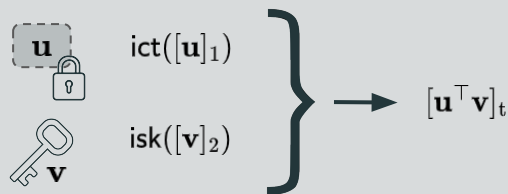
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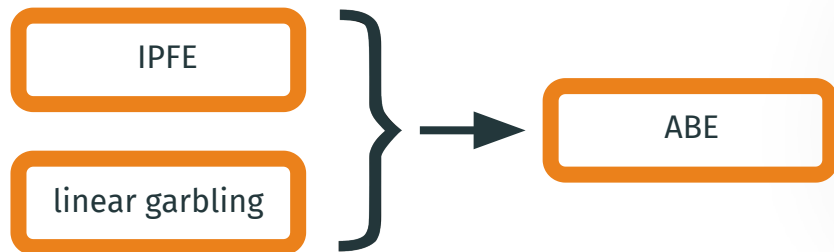
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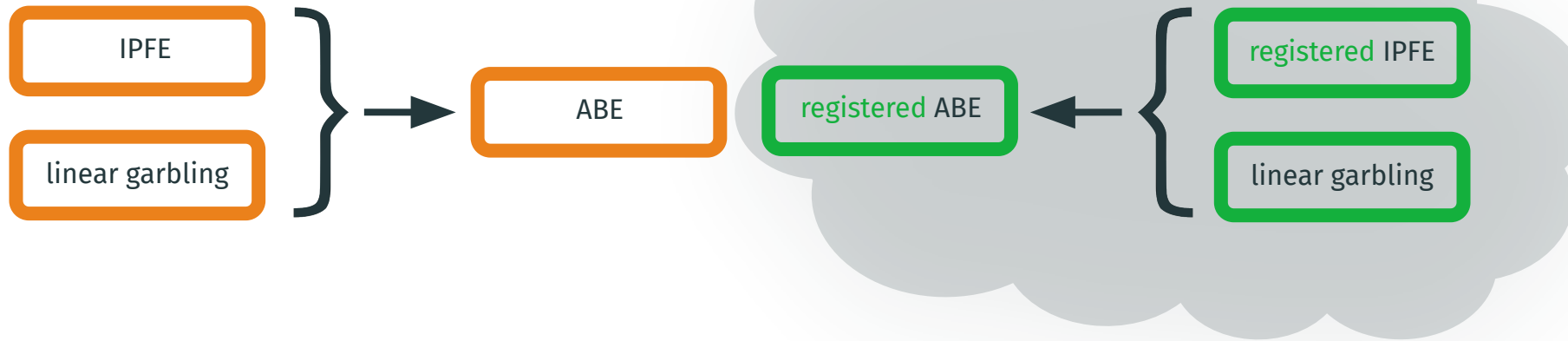
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## Linearity to the rescue.

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

-> offline/online phase



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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<sup>+</sup>24])

# How to pick the matrices?

## Arithmetic Key Garbling Scheme [EC:LL20]

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

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

Labels are

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

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

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*Security.*

*RIPFE leakage is*  $[\mathbf{d}_i]_{\mathbf{t}} = [\mathbf{u} \mathbf{P}_i \mathbf{V}_i]_{\mathbf{t}} = \left[ \left( \mu + s \overset{\$}{\sigma}_{i,0}, (\mathbf{x} \otimes (\overset{\$}{s} \sigma_{i,0}, \overset{\$}{s} \sigma_{i,1}, \overset{\$}{s} \mathbf{r}_i)) \cdot \hat{\mathbf{L}}_i \right) \right]_{\mathbf{t}} \xrightarrow{\text{indistinguishable from } \text{Sim}(f, x, d \leftarrow \mathcal{D})}$

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What about Turing machines?

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

$$[\mathbf{d}_i]_t = [\mathbf{u} \mathbf{P}_i \mathbf{V}_i]_t = \left[ \left( \mu + s \sigma_{i,0}, (\mathbf{x} \otimes (s \sigma_{i,0}, s \sigma_{i,1}, \mathbf{r}_i)) \cdot \hat{\mathbf{L}}_i \right) \right]_t$$

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- set of internal configurations  $\mathcal{C} = [N] \times [S] \times \{0, 1\}^S \times [Q]$  with initial configuration  $c_0 = (1, 1, \mathbf{0}_S, 1)$   
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*... so we only need to garble matrix multiplication*

# Arithmetic Key Garbling for Logspace TMs [EC:LL20]

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

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

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$$L_t[c, q](\mathbf{x}) = -(\mathbf{r}_{x,t-1} \otimes \mathbf{r}_f)[c, q] + ((\mathbf{r}_{x,t} \otimes \mathbf{r}_f) \cdot \mathbf{T}(\mathbf{x}))[c, q] = -\mathbf{r}_{x,t-1}[c] \cdot \mathbf{r}_f[q] + \underbrace{((\mathbf{r}_{x,t} \otimes \mathbf{r}_f) \cdot \mathbf{T}(\mathbf{x}))[c, q]}_{?}$$

$u$  (green arrow pointing to  $\mathbf{r}_{x,t-1}[c]$ )  
 $P_i$  (orange arrow pointing to  $\mathbf{r}_f[q]$ )

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- 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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Thank you!!! :)