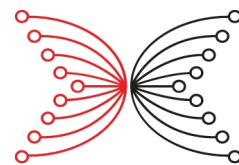


SNACKs: Leveraging Proofs of Sequential Work for Blockchain Light Clients

Hamza Abusalah¹, Georg Fuchsbauer¹, Peter Gaži², Karen Klein³



INPUT | OUTPUT

ETH zürich

Asiacrypt 2022 in Taipei, Taiwan



Outline

Light Client Blockchain Bootstrapping

Define SNACKs

(Succinct Non-interactive Argument of Chain Knowledge)

Construct SNACKs

(generically from Proofs of Sequential Work Schemes)

SNACKs to Bootstrap, generically

(Light-Client) Blockchain Protocols



(Light-Client) Blockchain Protocols



- sublinear effort (\mathcal{V})
- sublinear communication
- non-interactive

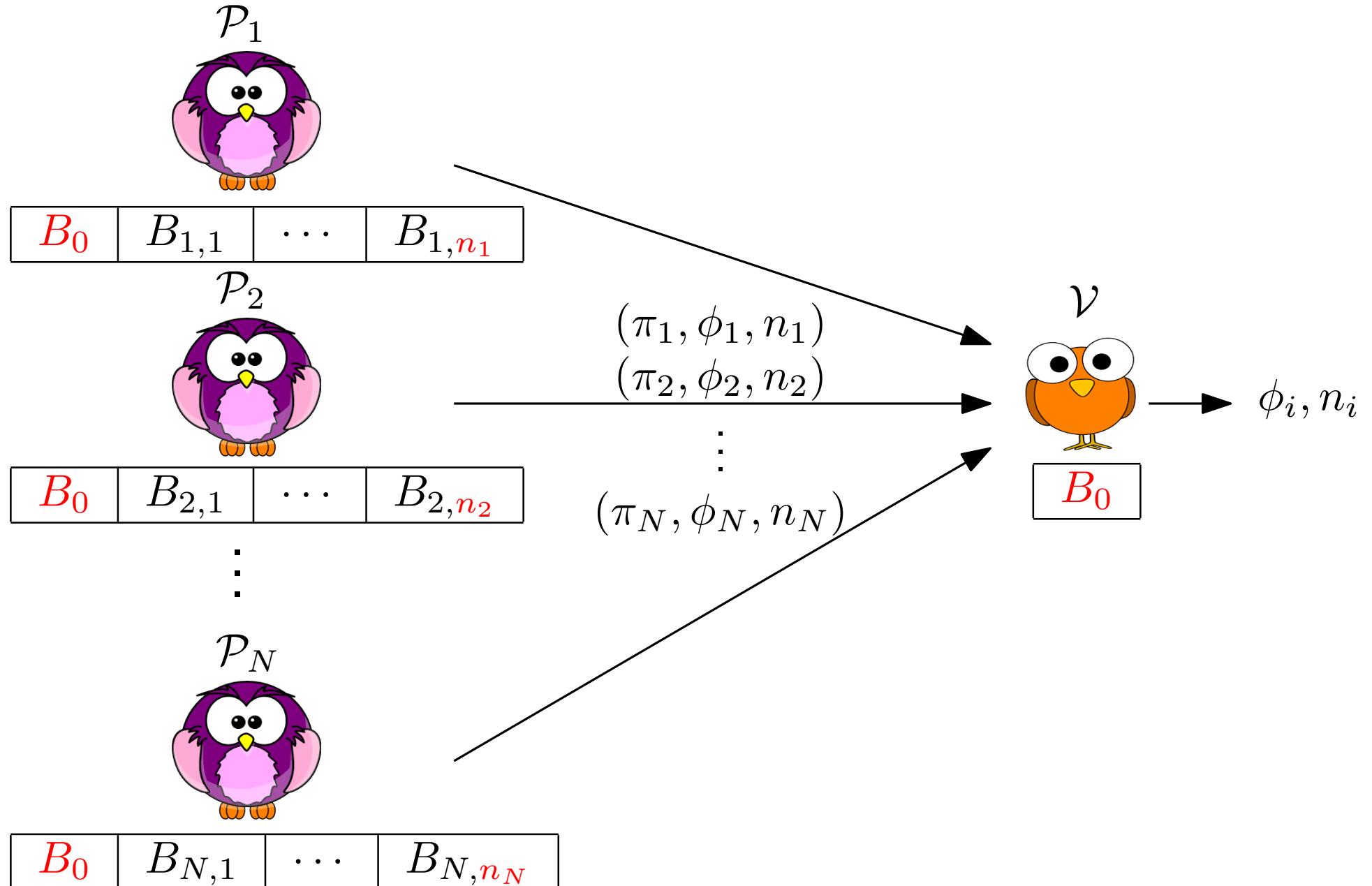
(Light-Client) Blockchain Protocols



- sublinear effort (\mathcal{V})
- sublinear communication
- non-interactive
- + sublinear effort (\mathcal{P})
- + no setup assumptions

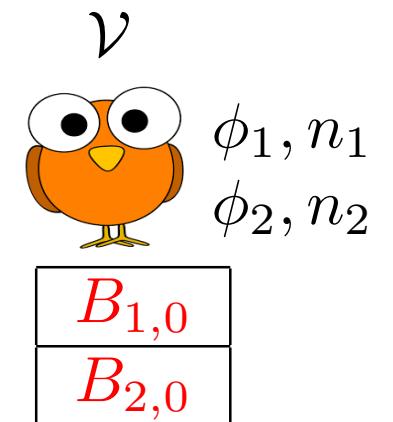
Bootstrapping

Ultimate goal of \mathcal{V} : a commitment to the stable prefix of the blockchain



Bootstrapped Verifiers

Bitcoin $\boxed{B_{1,0} \mid B_{1,1} \mid \dots \mid B_{1,n_1}}$



Chia $\boxed{B_{2,0} \mid B_{2,1} \mid \dots \mid B_{2,n_2}}$

Bootstrapped Verifiers

Bitcoin

$B_{1,0}$	$B_{1,1}$	\cdots	B_{1,n_1}
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\mathcal{P}
Statement over transactions in
 $B_{1,i}, B_{2,j}, \text{open}_i, \text{open}_j$

\mathcal{V}
Verifier ϕ_1, n_1
 ϕ_2, n_2

$B_{1,0}$
$B_{2,0}$

Chia

$B_{2,0}$	$B_{2,1}$	\cdots	B_{2,n_2}
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Solutions to Bootstrapping

- SNARKs
1), 3)
- Non-interactive proofs of proofs of work [Kiayias-Miller-Zindros 2020]
2), 3), 4)
 - Flyclient [Bünz-Kiffer-Luu-Zamani 2020]
2), 4)

Some limitations

- 1) Setup assumption
- 2) Limited applicability: PoW chains
- 3) Limited usefulness: no commitments
- 4) Non-standard models: **multiple** provers with ≥ 1 is **honest**

This Work: SNACKs

Define a **classical proof system** underlying all these applications:
Succinct Non-interactive Arguments of Chain Knowledge

Construct SNACKs from any Graph-Labeling Proof of Sequential Work

Unify and generalize GL-PoSW schemes, and provide a new construction

Application: realize bootstrapping from SNACKs

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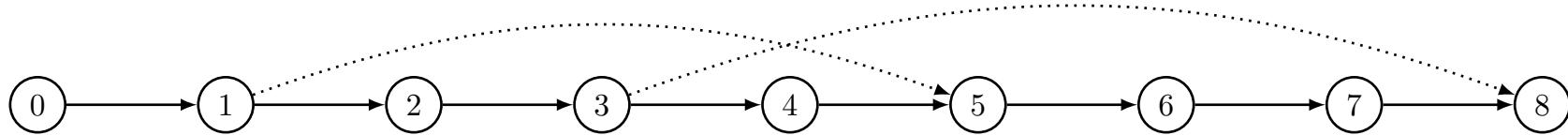
Application: realize bootstrapping from SNACKs

Advantages:

1. Classical soundness guarantees
2. Simplicity and modularity
3. Exchange of ideas between PoSW and Blockchains: *incremental* PoSW imply *incremental* SNACKs

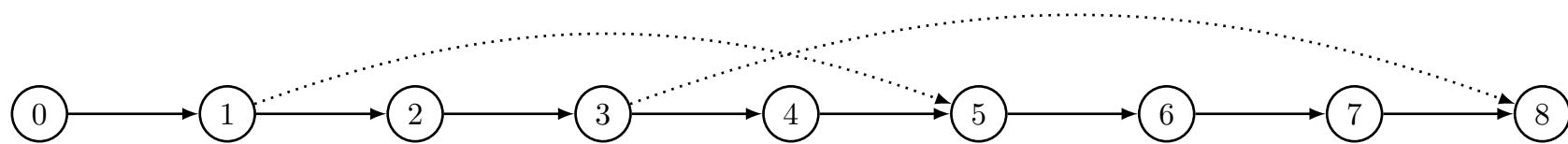
Constructing SNACKs

Blockchain DAG H with validity relation R_ψ s.t. $R_\psi(i, L(i), L(\text{paranets}(i))) = 1$ iff i th block is valid



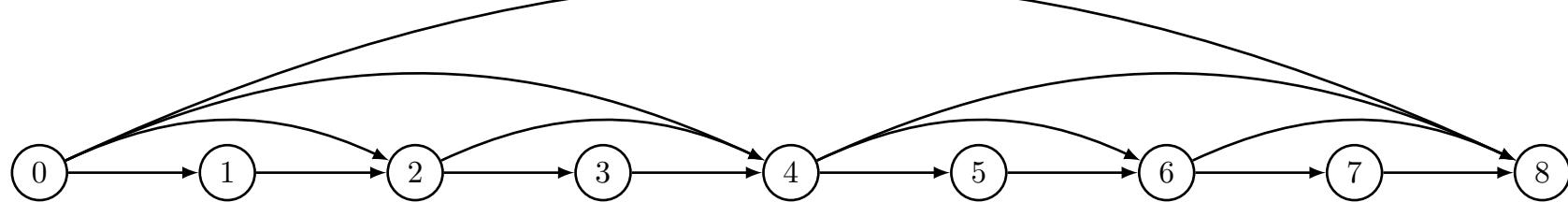
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PoS weighted DAG G



\mathcal{P}

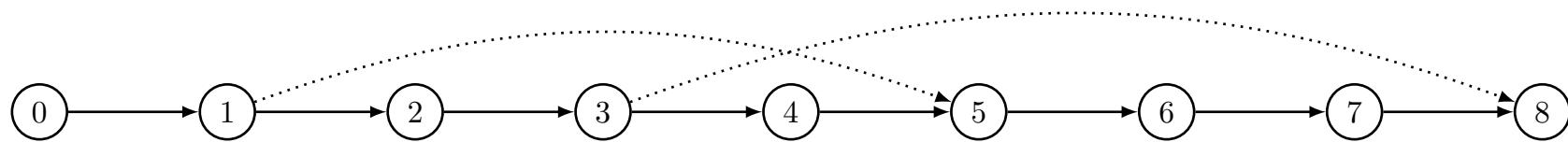


\mathcal{V}



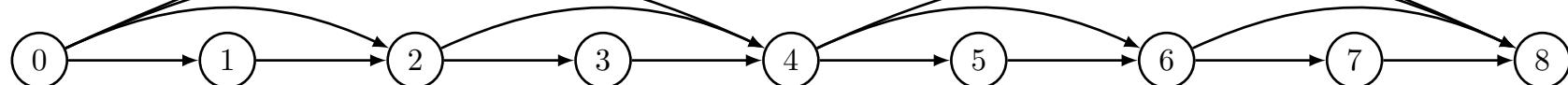
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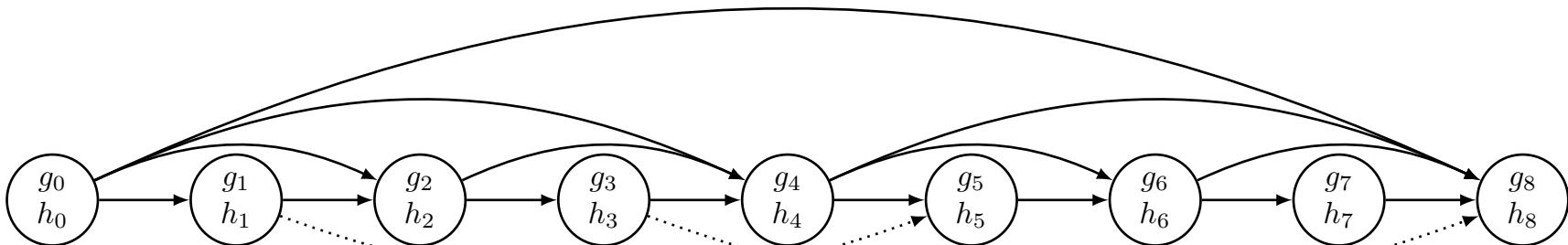
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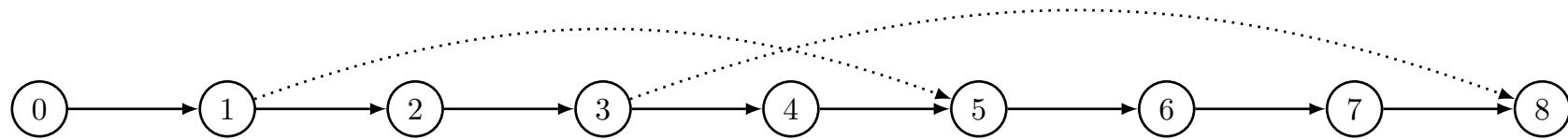
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Augmented weighted DAG $(K := ([n]_0, E_K := E_H \cup E_G), \Omega_n)$



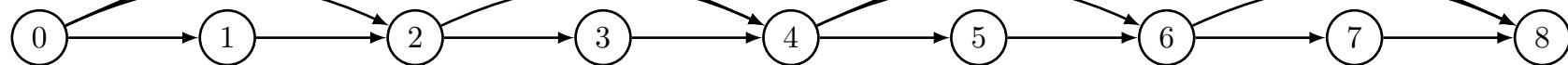
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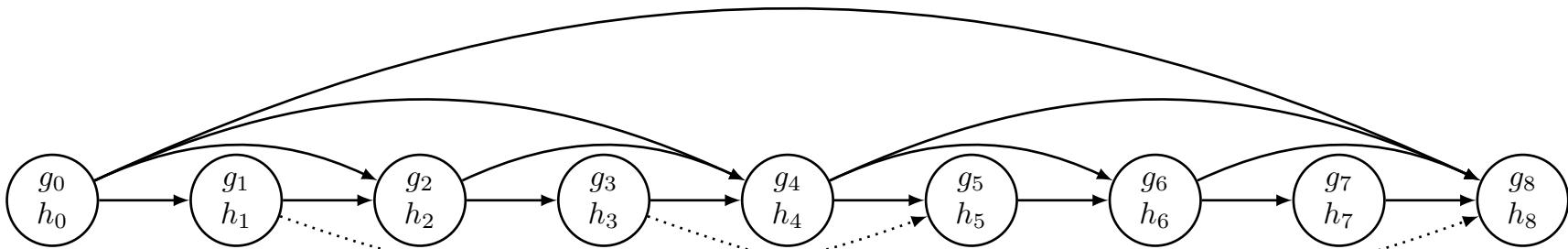
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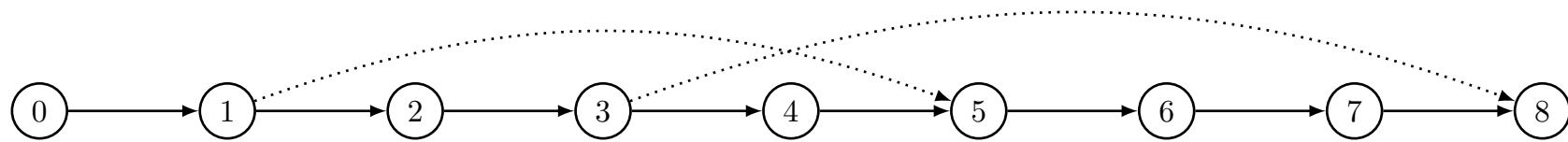
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$$\mathcal{R}^{(\alpha)} = \left\{ \left((\phi_n, n), (P, L_P, (w_i)_{i \in P}, \rho) \right) : \begin{array}{l} (P, L_P, (w_i)_{i \in P}) \text{ is } (\alpha, R)\text{-valid} \\ \wedge \text{Comm.ver}(\phi_n, L_P, P, \rho) = 1 \end{array} \right\}$$

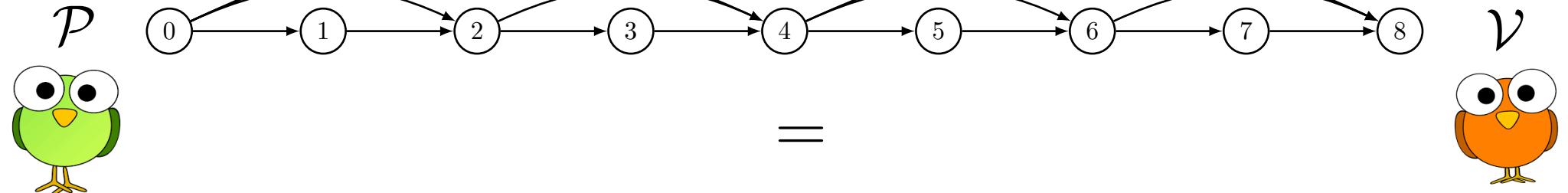
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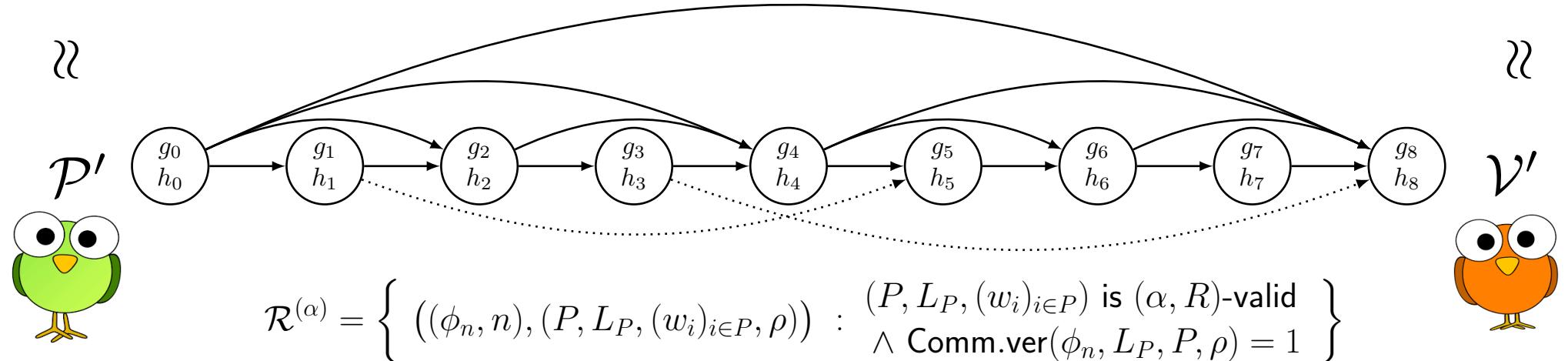


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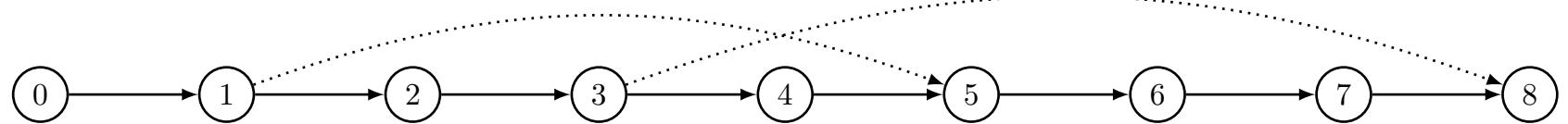


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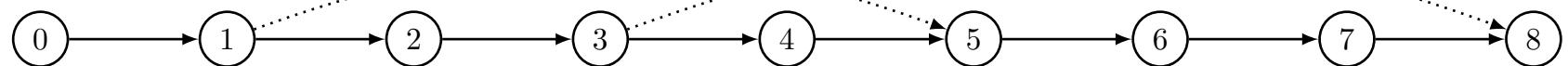
DAG Labeling

Chain: $C_n = ([n]_0, E_n)$ a DAG with $E_n \supseteq \{(i - 1, i) : i \in [n]\}$



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τ -based graph labeling: $\tau : \{0, 1\}^* \rightarrow \{0, 1\}^\lambda$ (modeled as a RO)

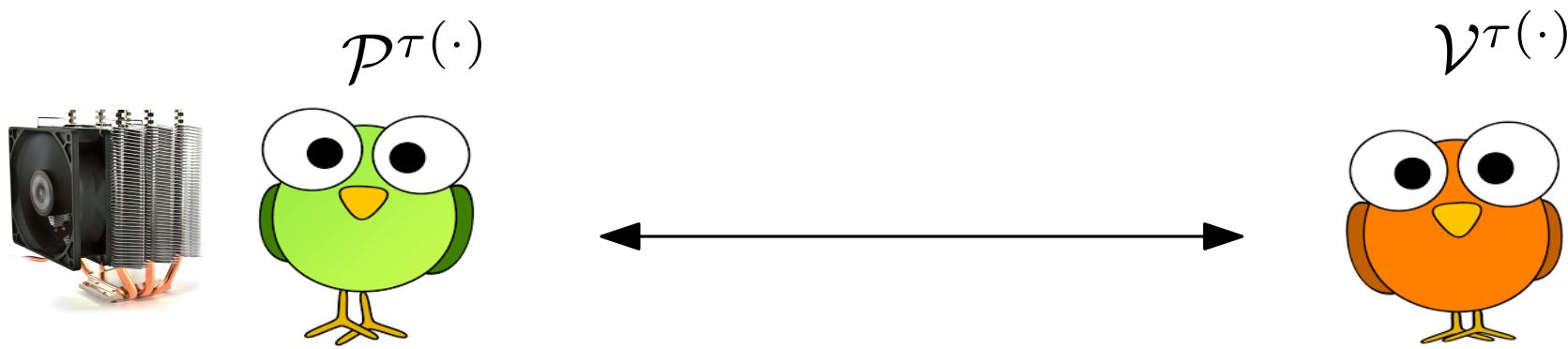
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$$E.g., L(5) = \tau(5, L(1), L(4))$$

Proof of Sequential Work

[Mahmoody-Moran-Vadhan 2013]

Parameters: DAG G_n, \dots



Completeness: Honest $\mathcal{P}^{\tau(\cdot)}$ making n **sequential** $\tau(\cdot)$ queries makes \mathcal{V} accept w.p. 1

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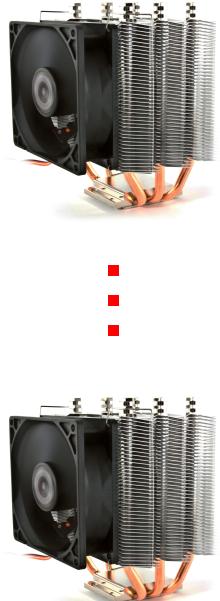
Succinctness: For every honest proof π :

$|\pi| \leq \text{poly}(\lambda, \log n)$, $\text{Time}(\mathcal{V}) \leq \text{poly}(\lambda, \log n)$, and $\text{Time}(\mathcal{P}) \leq \text{poly}(\lambda, n)$

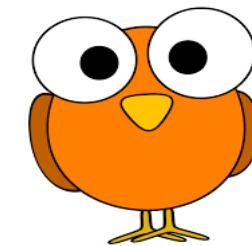
Proof of Sequential Work

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$\mathcal{V}^\tau(\cdot)$



$\text{poly}(\lambda, \log n)$ time

(α, ϵ) -Soundness: A parallel $\tilde{\mathcal{P}}^\tau(\cdot)$ making $\leq \alpha \cdot n$ **sequential** queries to $\tau(\cdot)$ fails with prob. $\geq 1 - \epsilon(\lambda)$

A Simple PoSW

This work

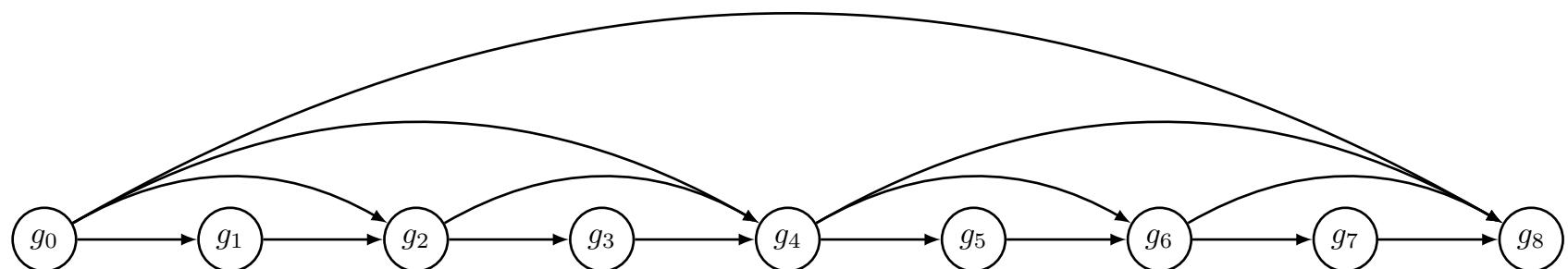
Parameters: G_n, \dots



$\mathcal{P}^{\mathcal{O}(\cdot)}$



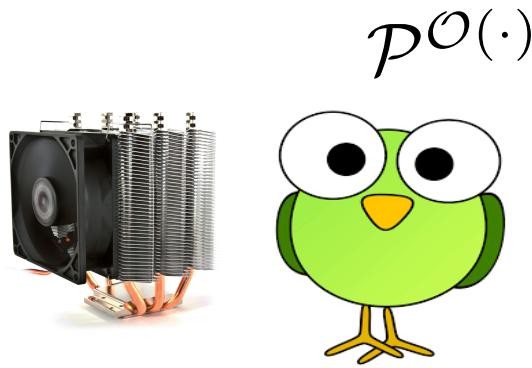
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A Simple PoSW

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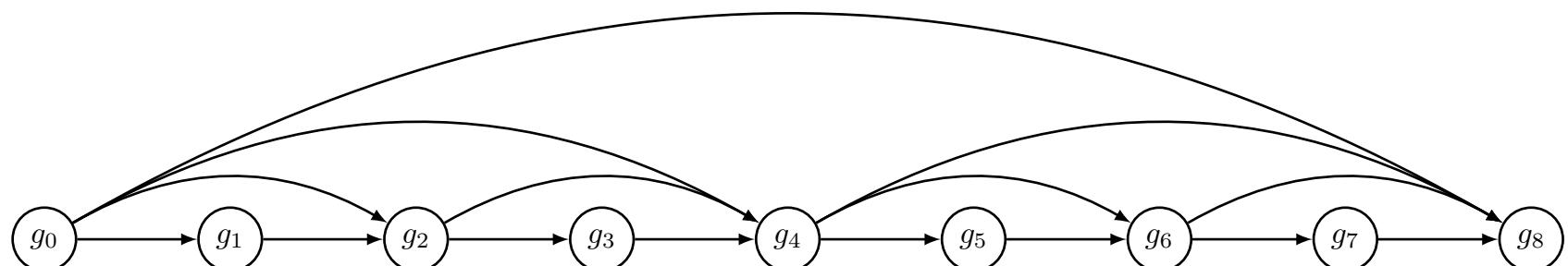
$$\chi$$

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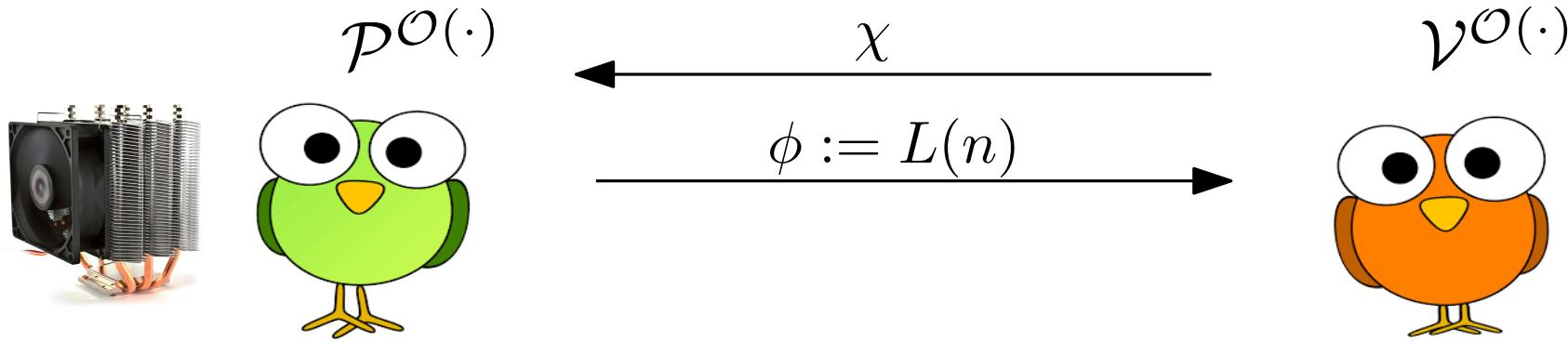
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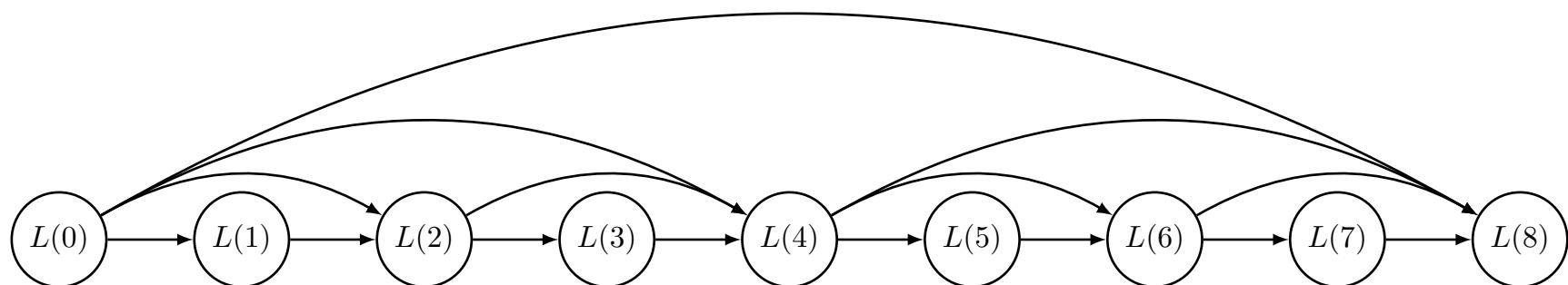
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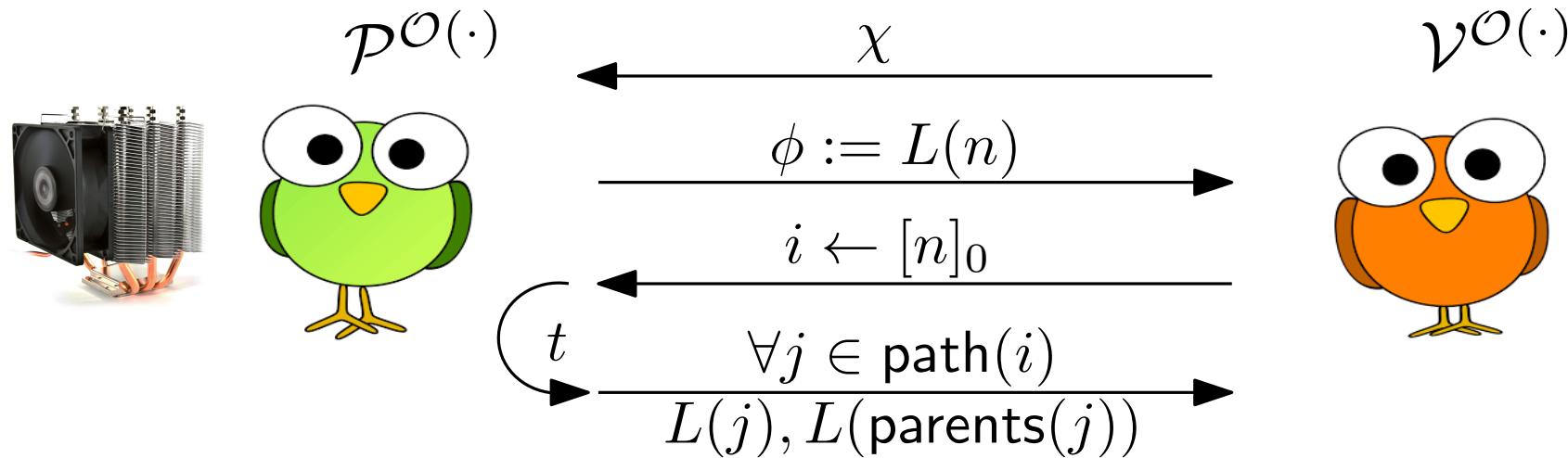
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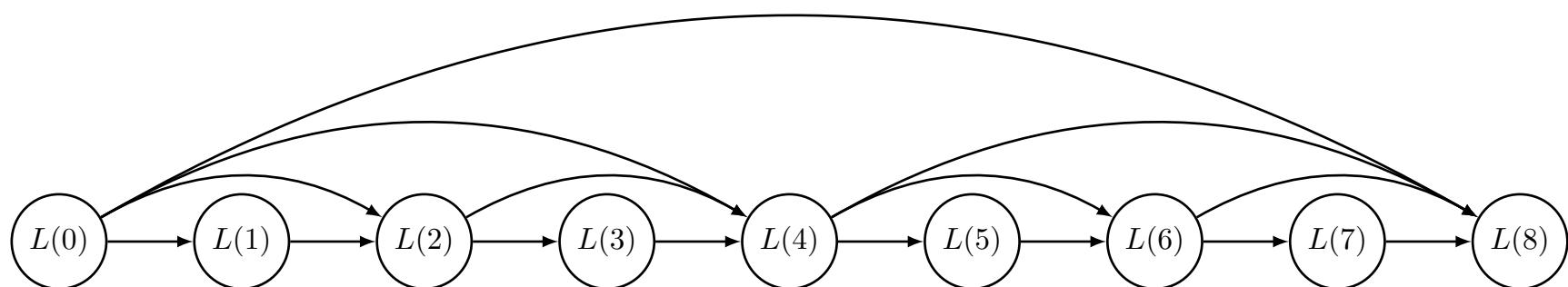
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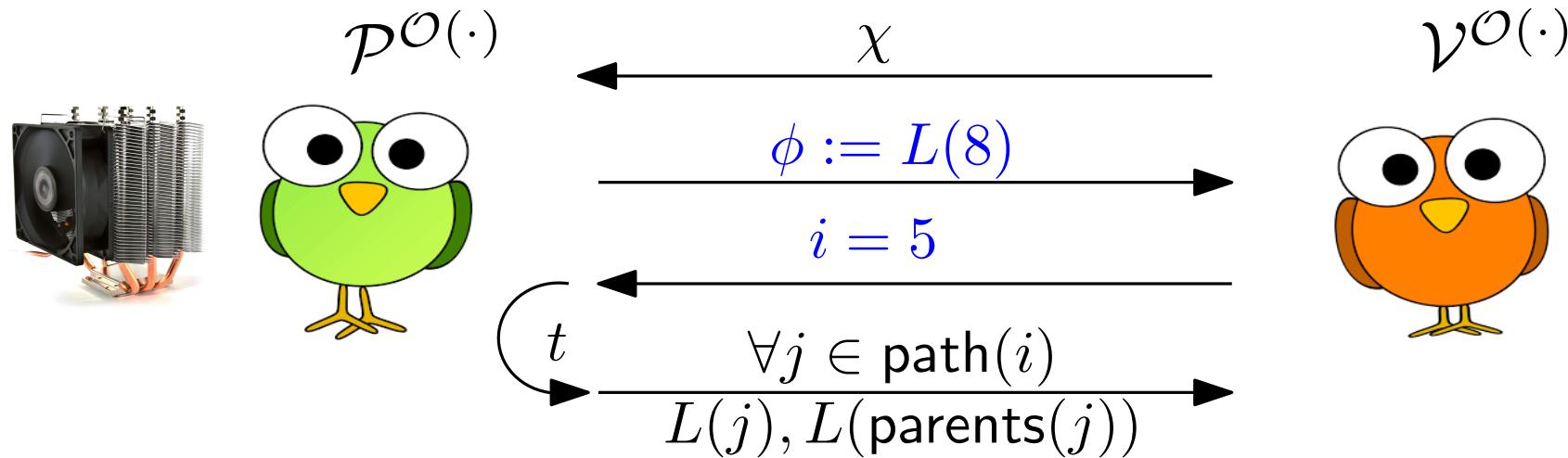
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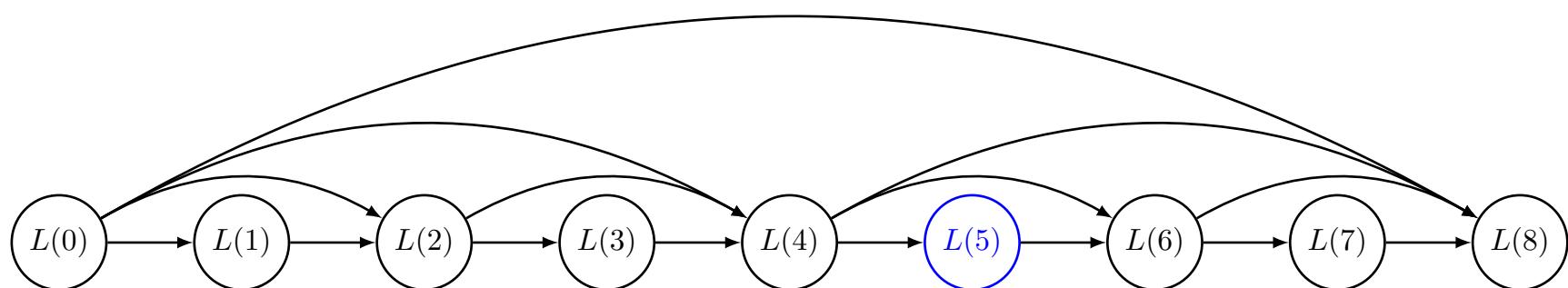
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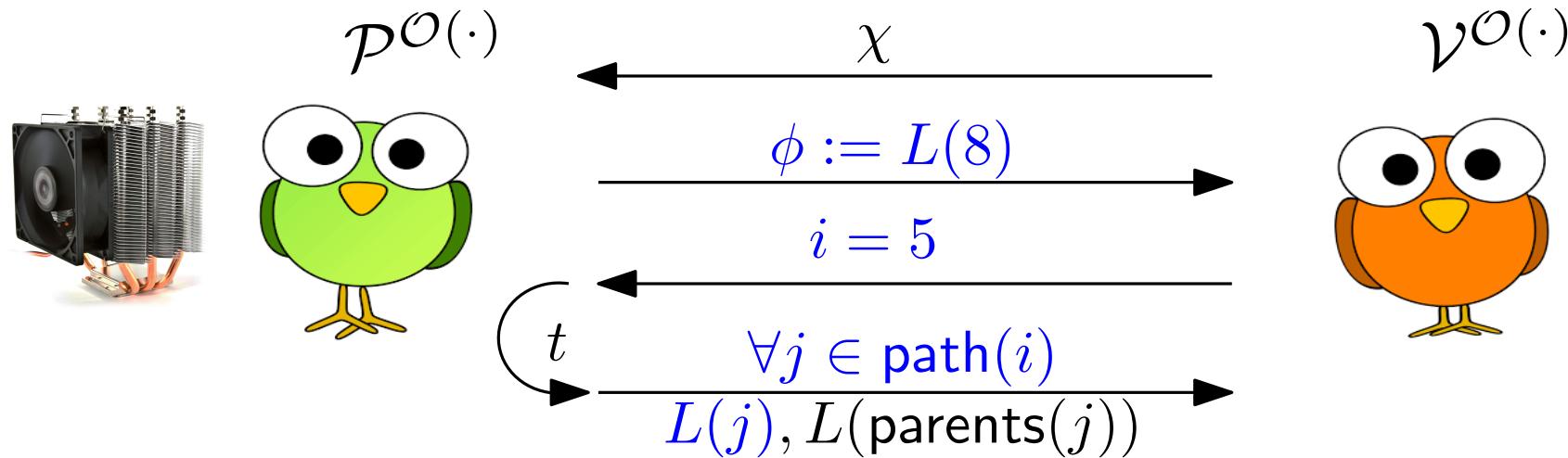
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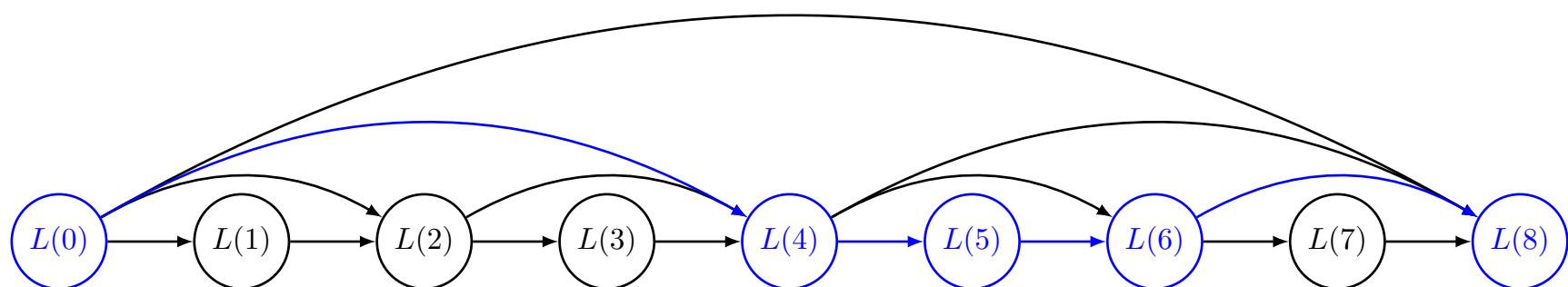
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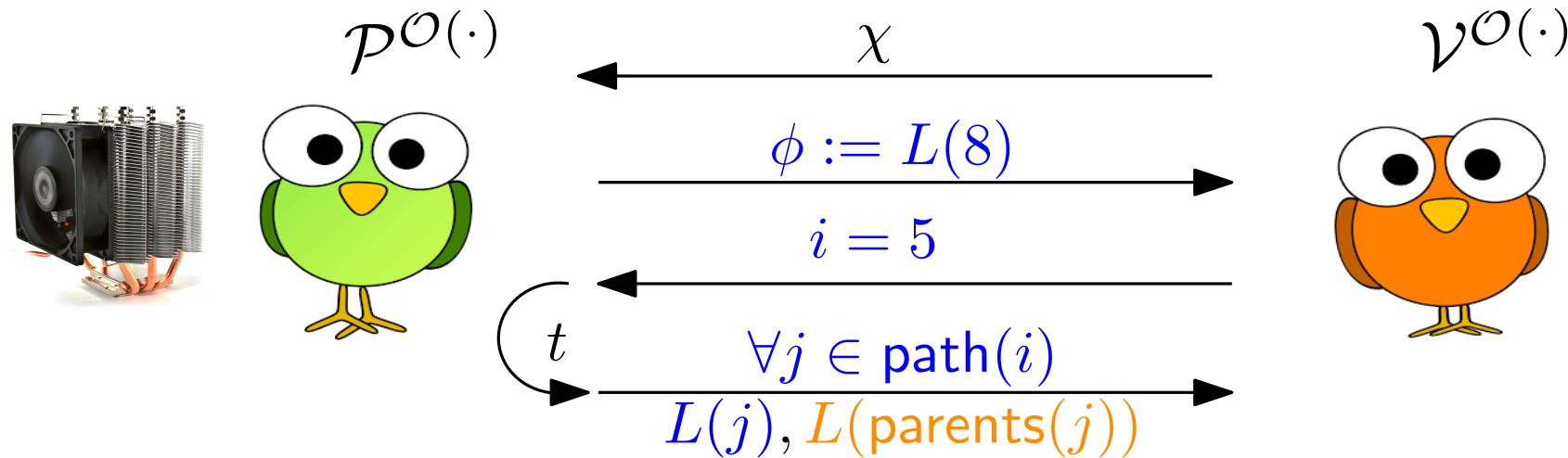
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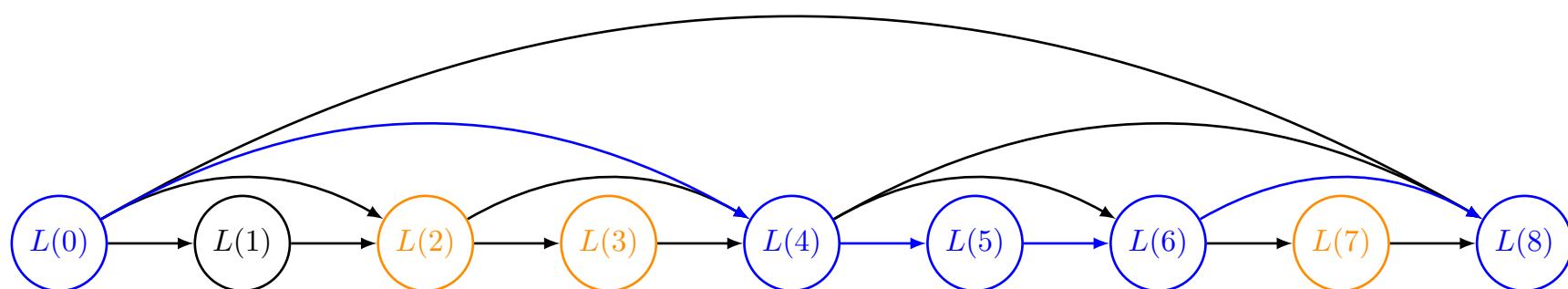
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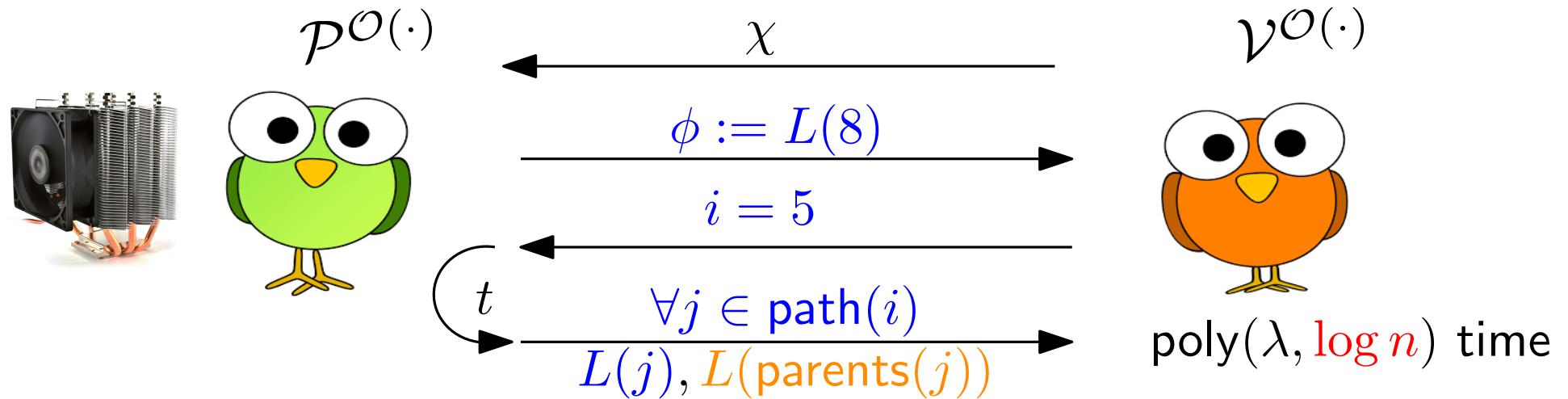
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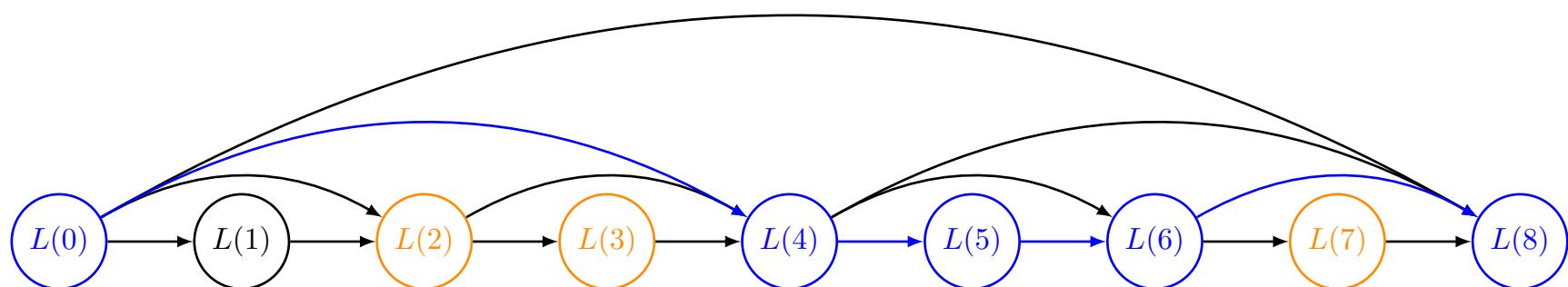
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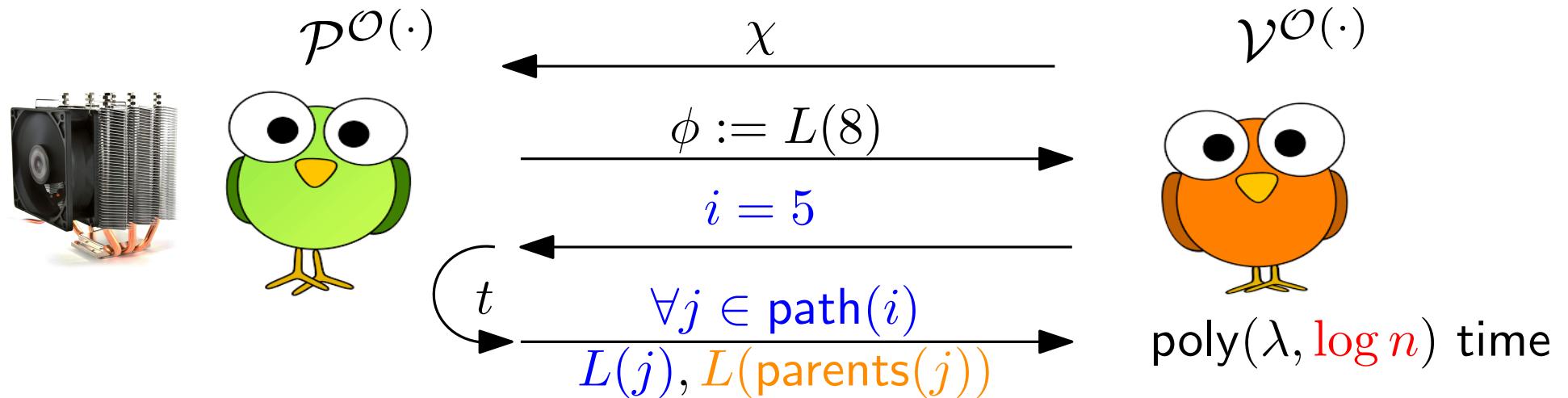
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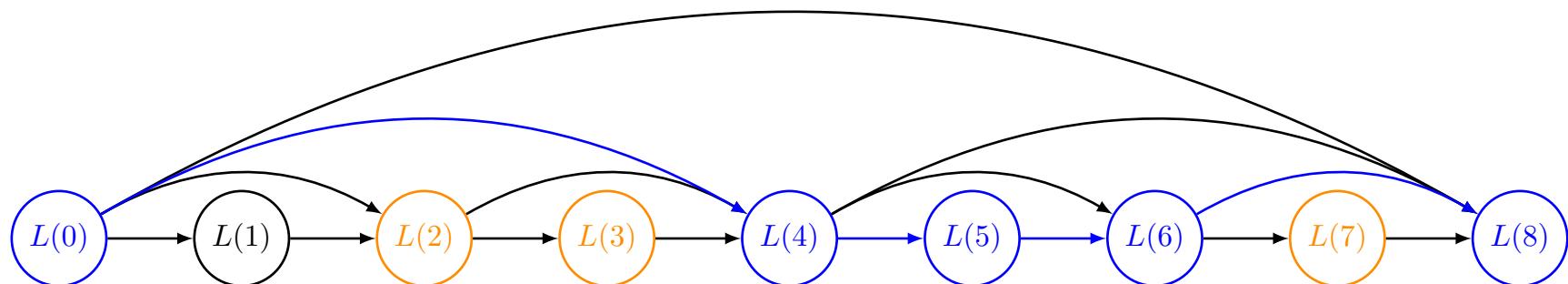
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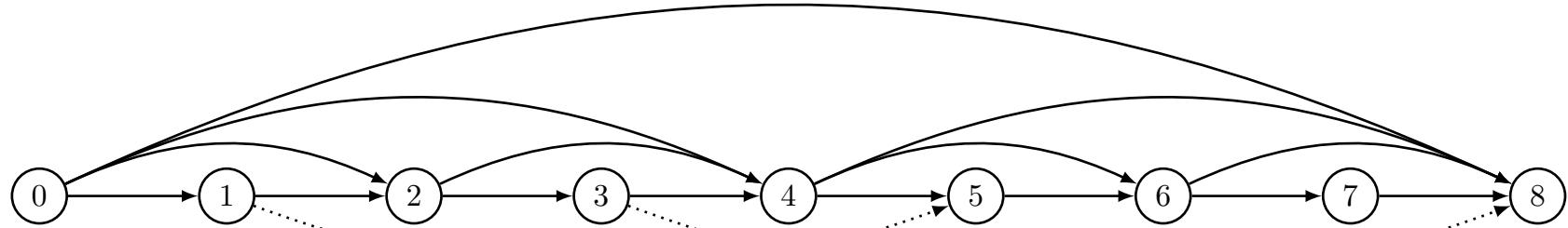
Thm: If 1. $\tilde{\mathcal{P}}_1$ made $\leq \alpha \cdot n$ sequential queries to $\tau(\cdot)$ before sending ϕ
 2. $\tilde{\mathcal{P}} := (\tilde{\mathcal{P}}_1, \tilde{\mathcal{P}}_2)$ made a total of $\leq q$ queries to $\tau(\cdot)$
 Then $\tilde{\mathcal{P}}$ makes \mathcal{V} accept with probability at most

$$\epsilon := \alpha^t + 3 \cdot q^2 / 2^\lambda$$



Towards Defining SNACKs

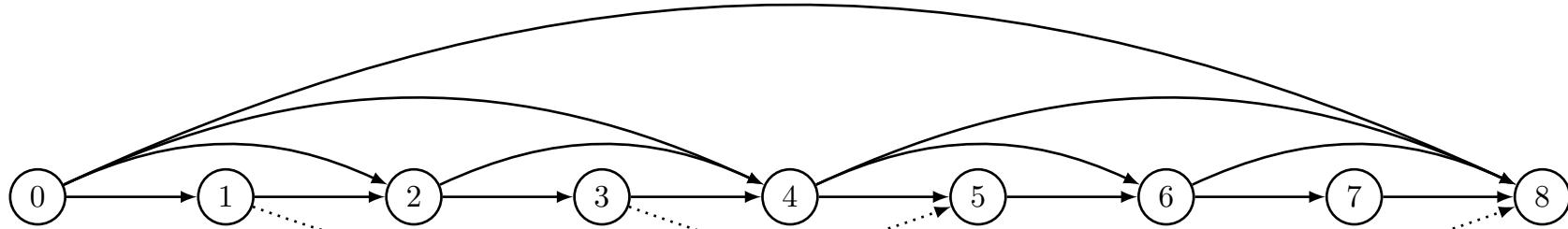
Weighted: $\Gamma_n = (C_n, \Omega_n)$ with $\Omega_n : [n]_0 \rightarrow [0, 1]$ s.t. $\sum_{i=0}^n \Omega_n(i) = 1$



$$\Omega_n(0) + \Omega_n(1) + \Omega_n(2) + \Omega_n(3) + \Omega_n(4) + \Omega_n(5) + \Omega_n(6) + \Omega_n(7) + \Omega_n(8) = 1$$

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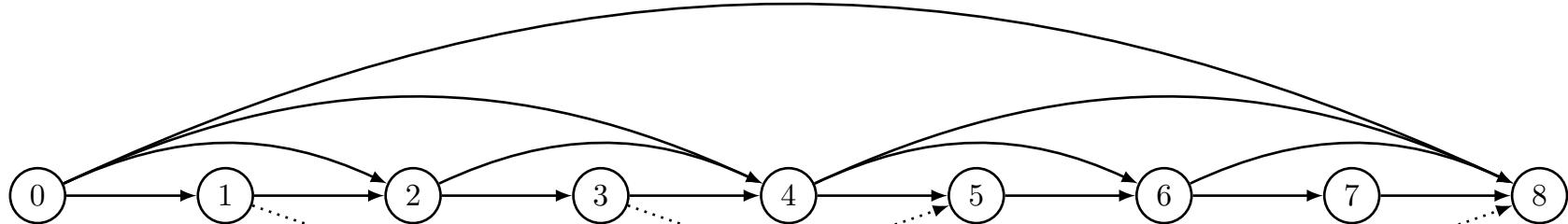
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(x_0, \dots, x_n) -augmented τ -based graph labeling: $\tau : \{0, 1\}^* \rightarrow \{0, 1\}^\lambda$

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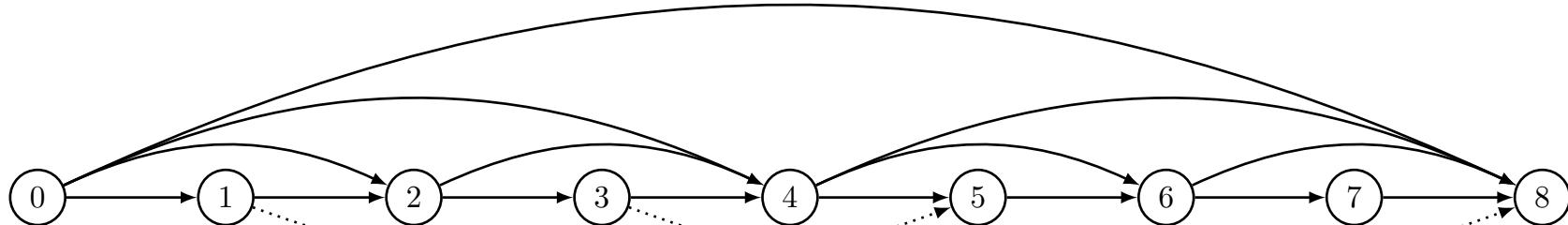
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Validity of augmented labels via $R \subseteq \mathbb{N}_0 \times (\{0, 1\})^2$:

$$R(i, L(i), w_i) = 1 \text{ iff } L(i) = \tau(i, w_i) \| x'_i \quad \wedge \quad R_\psi(i, L(i), w_i) = 1$$

Towards Defining SNACKs

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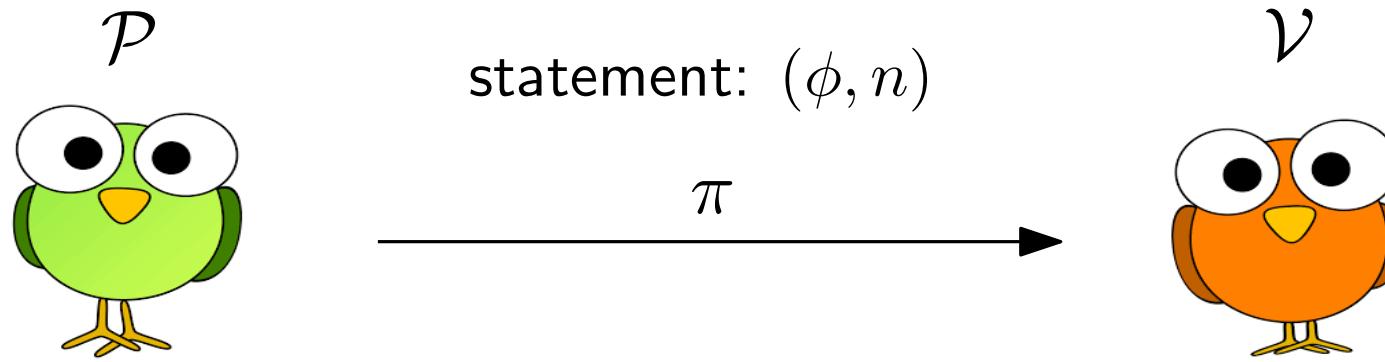
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Valid path: $(P, L_P, (w_i)_{i \in P})$ in Γ_n for R is (α, R) -valid path if

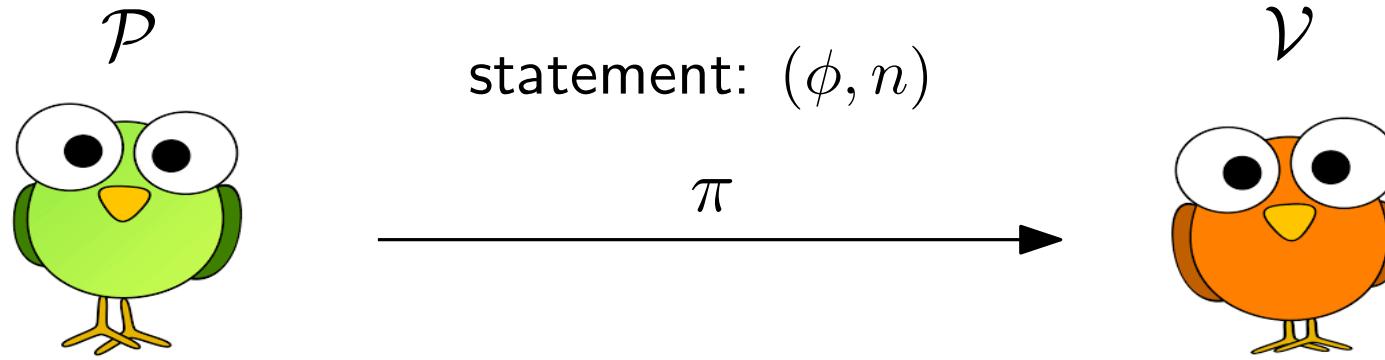
$$\forall i \in P : R(i, L_P(i), w_i) = 1 \quad \text{and} \quad \Omega_n(P) \geq \alpha$$

Defining SNACKs



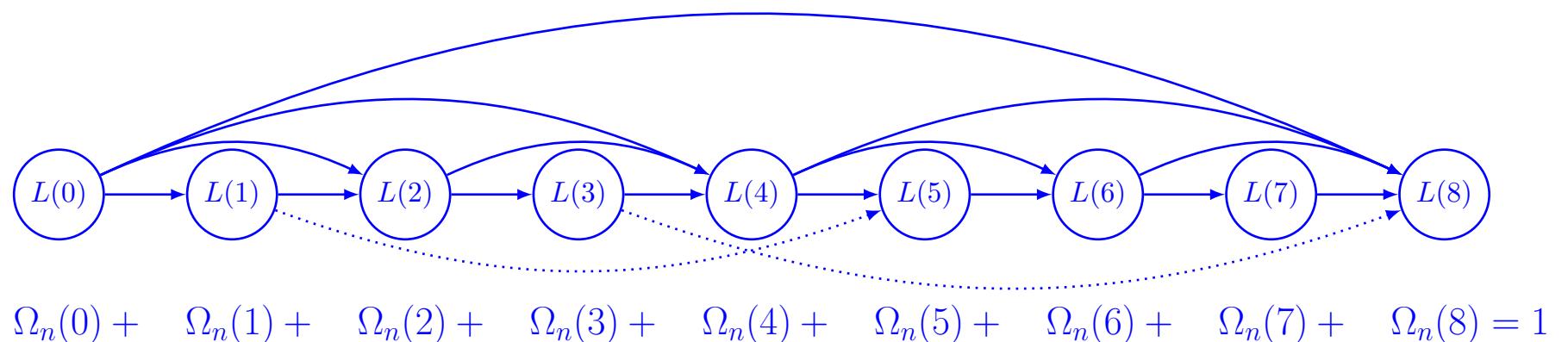
$$\mathcal{R}^{(\alpha)} = \left\{ \begin{array}{l} ((\phi, n), \\ (P, L_P, (w_i)_{i \in P}, \rho)) \end{array} : \begin{array}{l} (P, L_P, (w_i)_{i \in P}) \text{ is } (\alpha, R)\text{-valid} \\ \wedge \text{Comm.ver}(\phi, L_P, P, \rho) = 1 \end{array} \right\}$$

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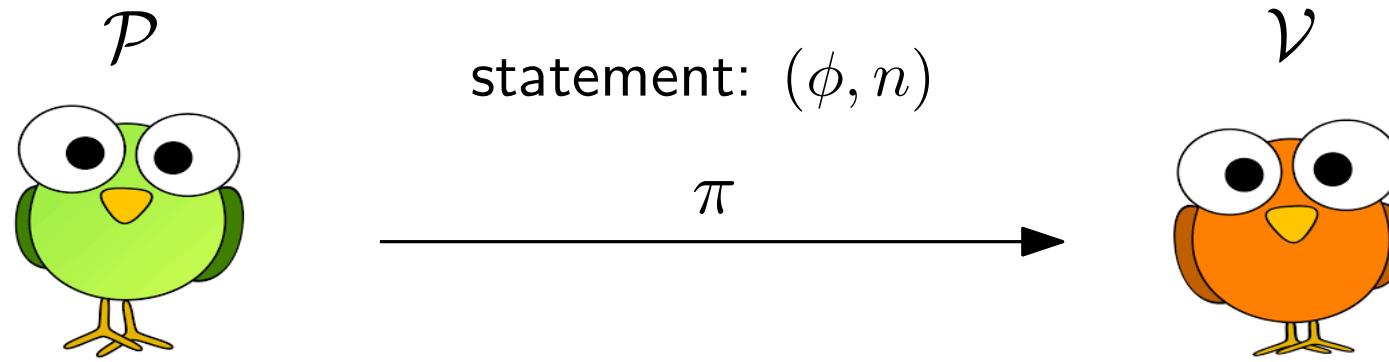


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Completeness: For every $((\phi, n), s) \in \mathcal{R}^{(\alpha=1)}$, \mathcal{V} accepts w.p. 1

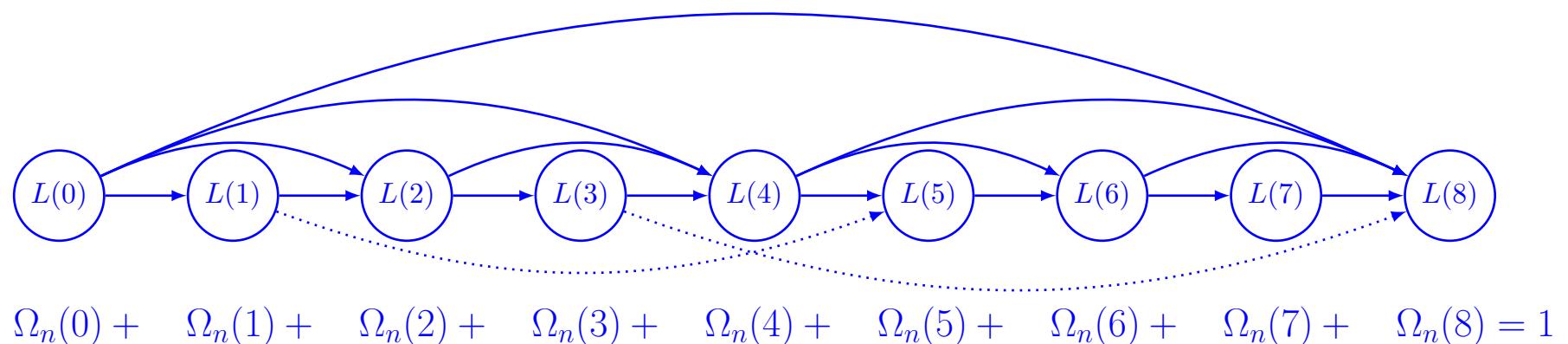


Defining SNACKs



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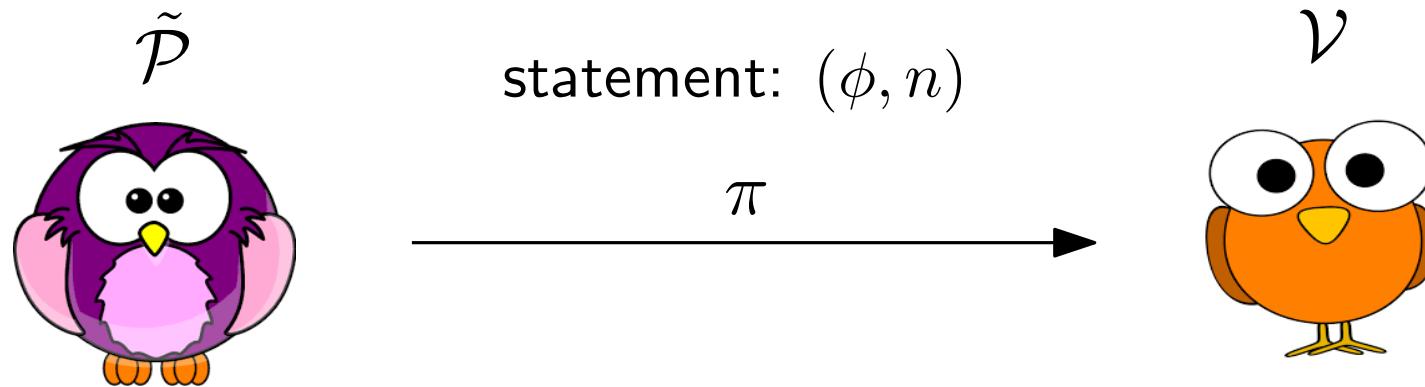
Completeness: For every $((\phi, n), s) \in \mathcal{R}^{(\alpha=1)}$, \mathcal{V} accepts w.p. 1



Succinctness: For every honest π

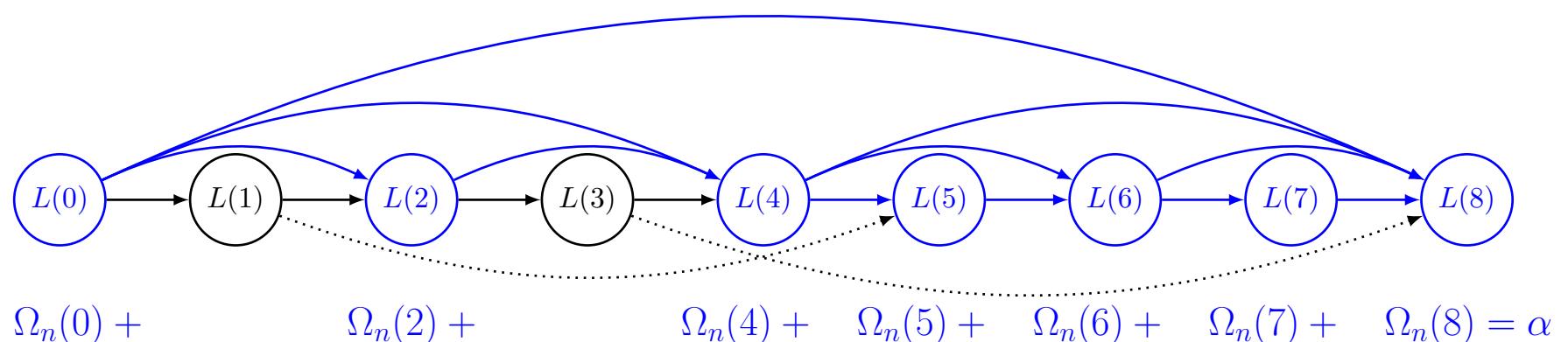
$|\pi| \leq \text{poly}(\lambda, \log n)$, $\text{Time}(\mathcal{V}) \leq \text{poly}(\lambda, \log n)$, and $\text{Time}(\mathcal{P}) \leq \text{poly}(\lambda, n)$

Defining SNACKs



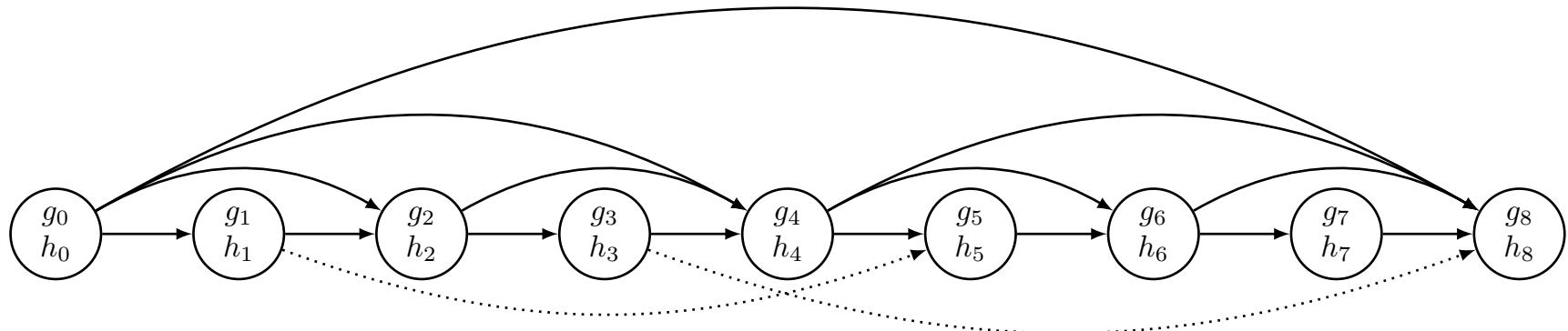
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(α, ϵ) -Knowledge Soundness: $\forall \tilde{\mathcal{P}}$, there exists an efficient \mathcal{E} that extracts witness s s.t. $((\phi, n), s) \in \mathcal{R}^{(\alpha)}$ w.p. $\geq 1 - \epsilon(\lambda)$



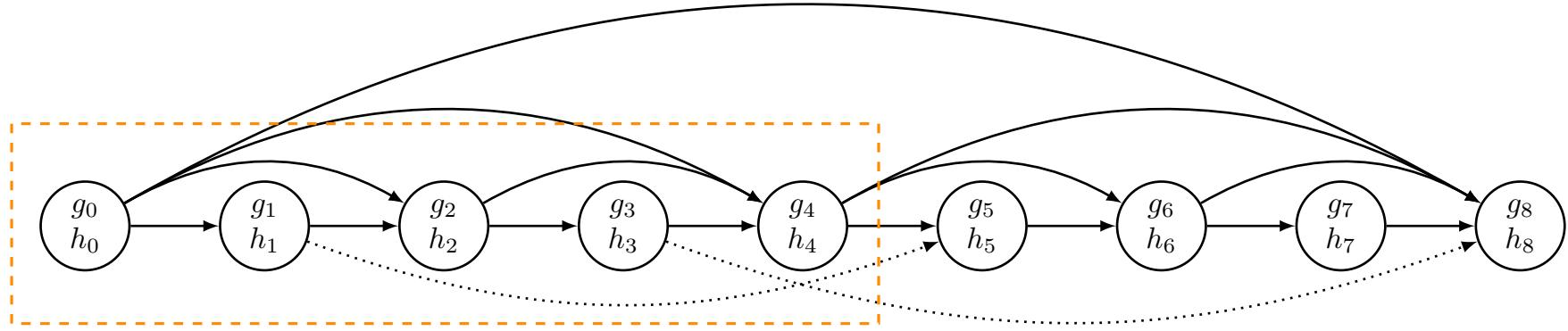
Labeling the PoSW-Augmented Blockchain

Augmented weighted DAG $\Gamma_n := (K := ([n]_0, E_K := E_H \cup E_G), \Omega_n)$



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Algorithm Init:

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1. $\chi \leftarrow \{0,1\}^\lambda$
2. $\ell_0 := \tau_0(\varepsilon)$
3. $(\phi_0, \text{aux}_0) \leftarrow \text{Com.commit}(\ell_0)$
4. $g_0 := \ell_0 \parallel \phi_0$
5. $h_0 := (0, d_0 := \psi \parallel \chi \parallel \pi_0 := \varepsilon)$
6. **return** $(\psi := L_K(0) := k_0 := (g_0, h_0), \text{aux}_0)$

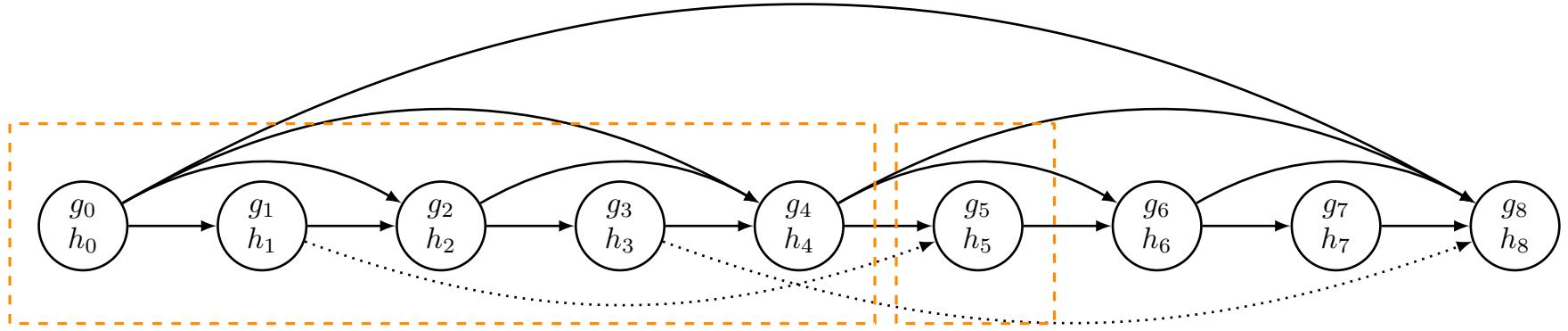
Algorithm Mine:

On input $((k_j := (g_j, h_j))_{j \in [i-1]_0}, \text{data}_i)$:

1. $\ell_i := \tau(i, L_K(\text{parents}_K(i)))$
2. $(\phi_i, \text{aux}_i) \leftarrow \text{Com.commit}((k_j)_{j \in [i-1]_0} \parallel \ell_i)$
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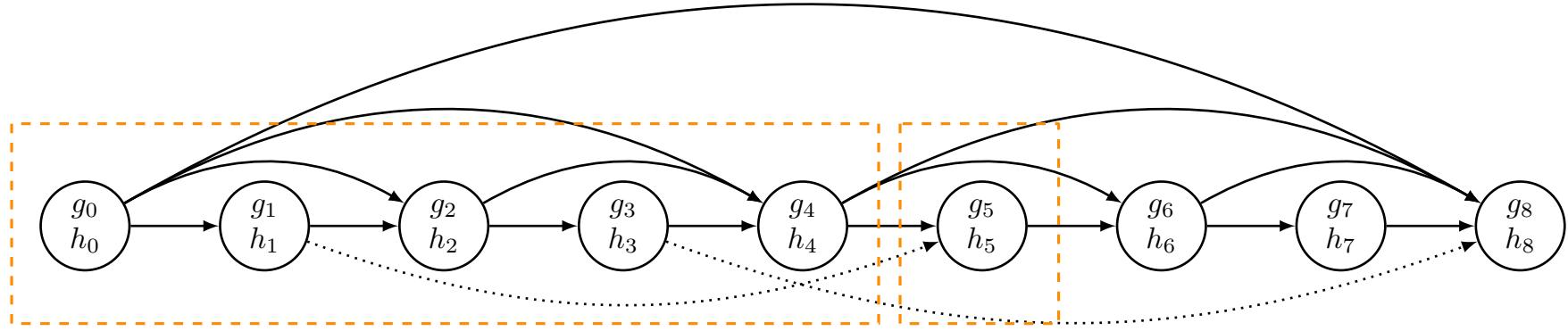
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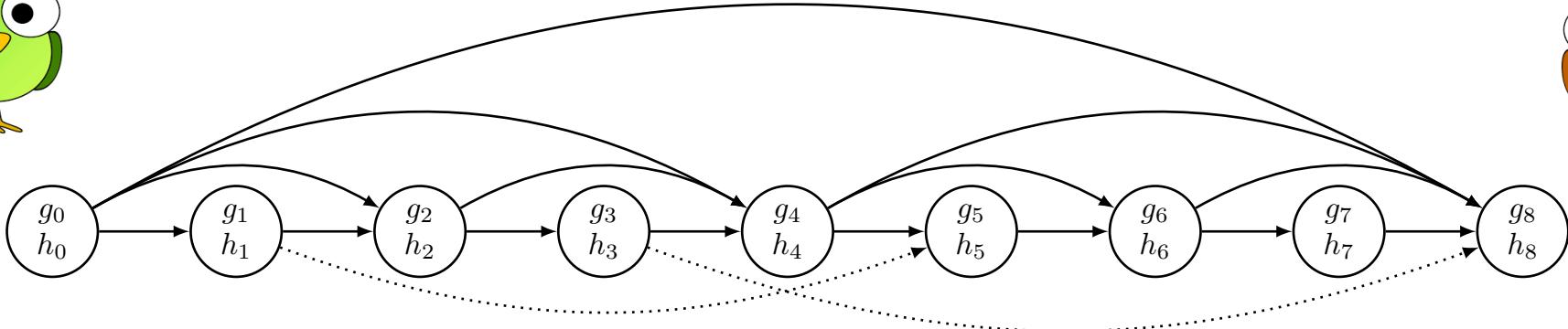
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2. $(\phi_i, \text{aux}_i) \leftarrow \text{Com.commit}((k_j)_{j \in [i-1]_0} \parallel \ell_i)$
for this PoSW constr.: $\phi_i := \ell_i$
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SNACK

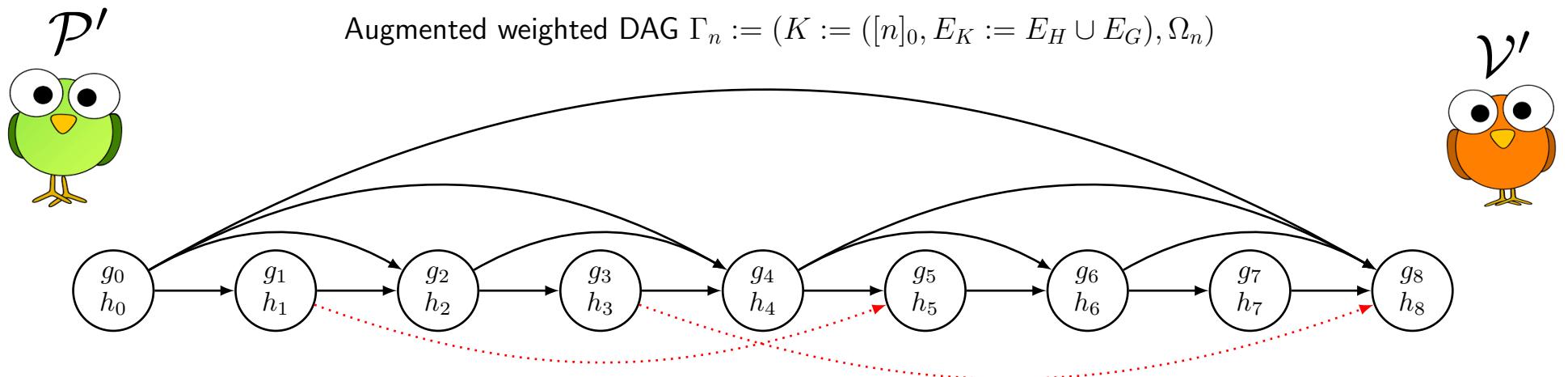


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SNACK

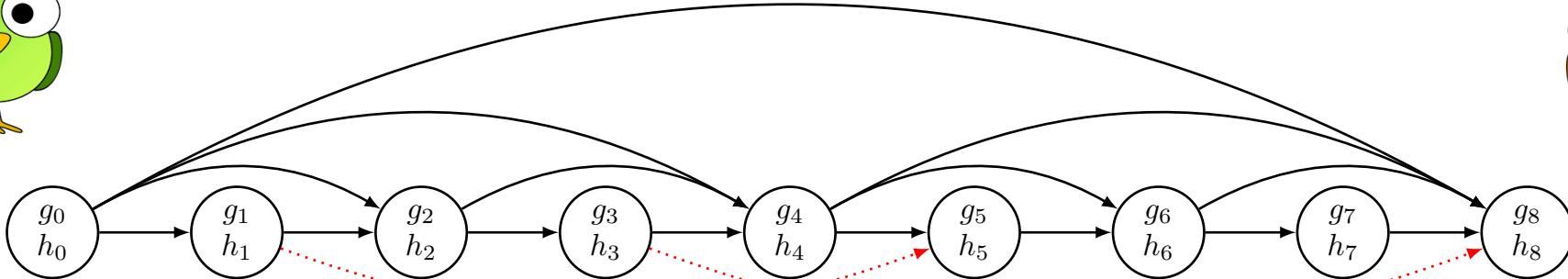


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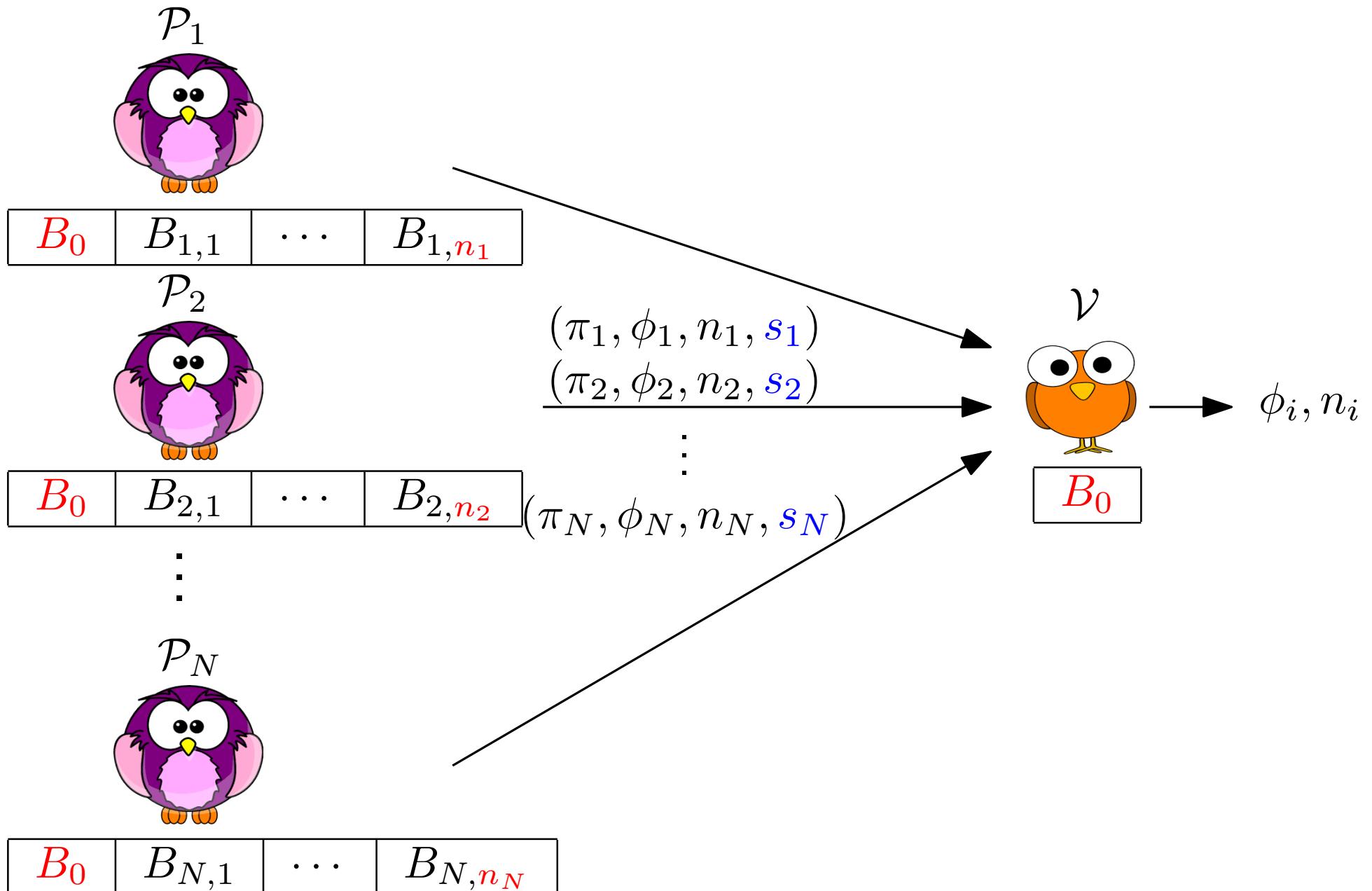
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Thm: (α, ϵ) -knowledge soundness of SNACK follows directly from any (α, ϵ) -knowledge soundness of GL-PoSW.

(after applying Fiat-Shamir)

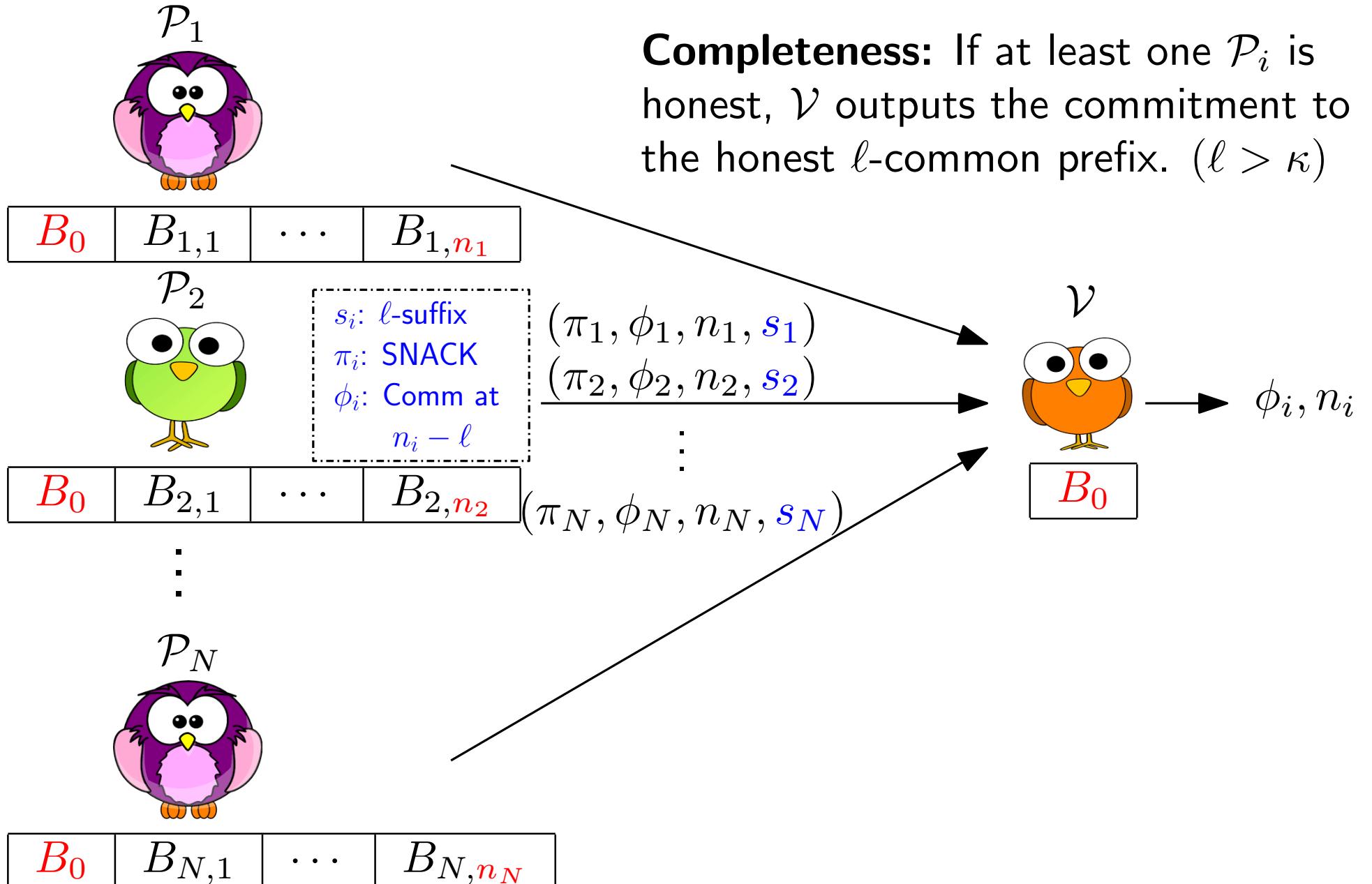
SNACK Bootstrapping

Goal: \mathcal{V} holds a commitment to a sufficiently-long stable prefix



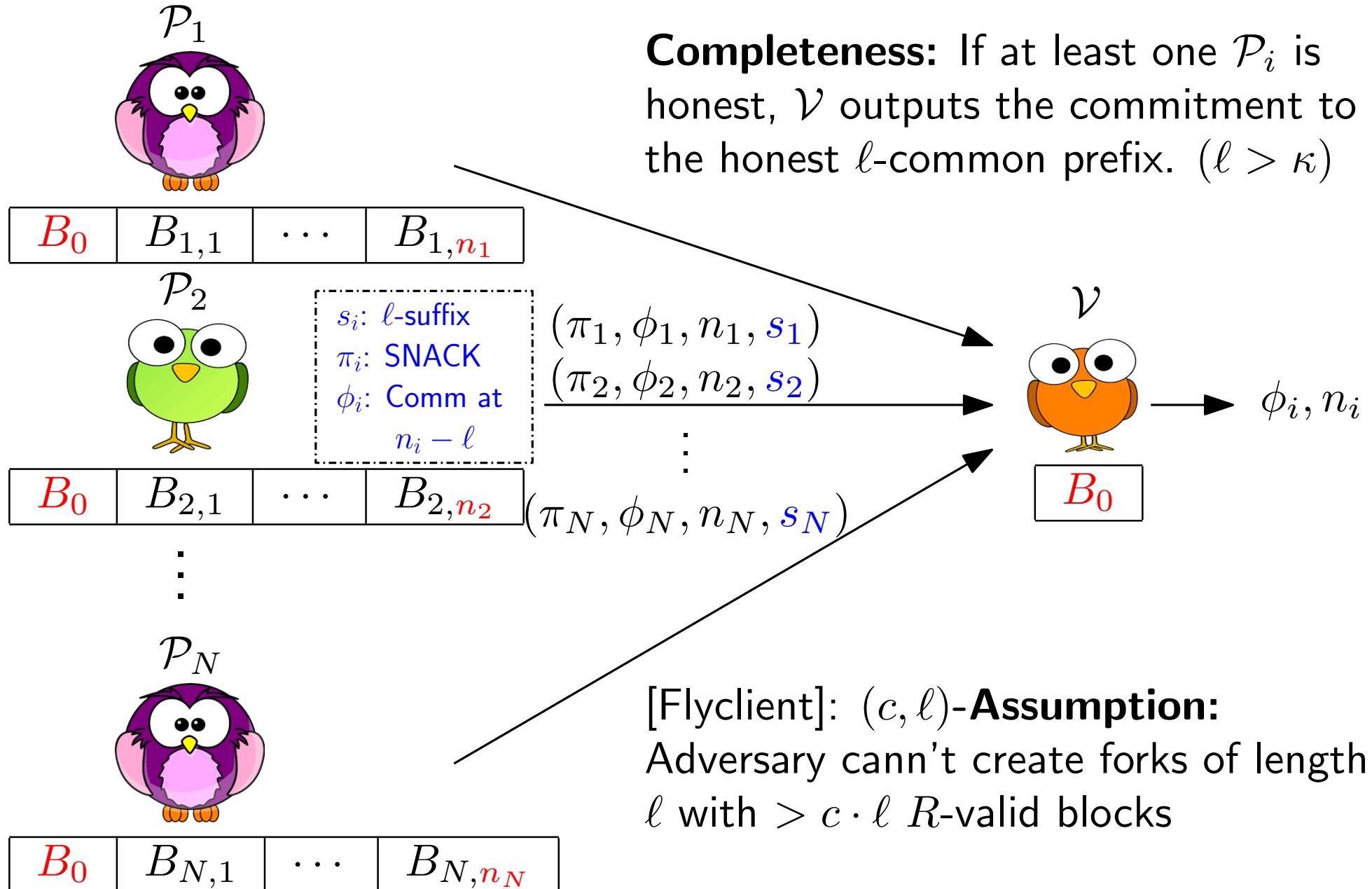
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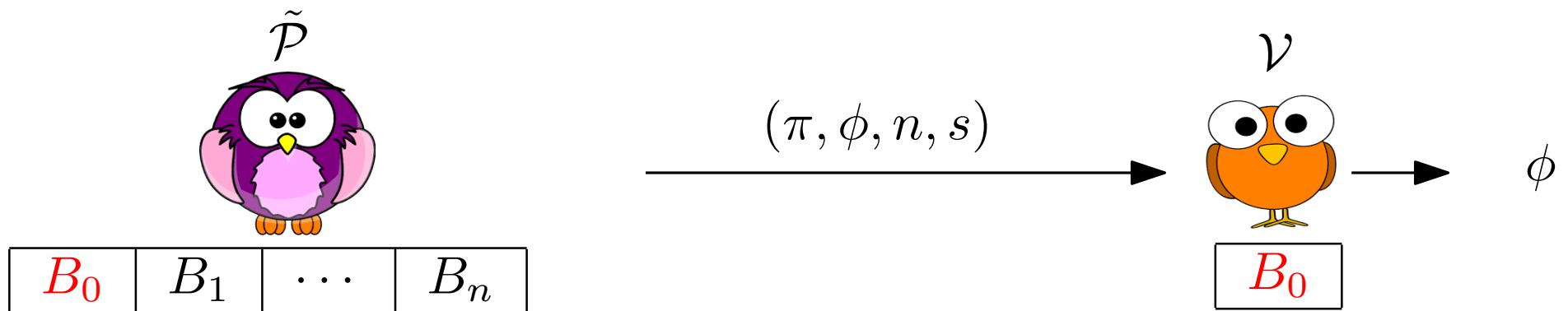
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SNACK Bootstrapping

(α_n, ϵ) -Knowledge-Soundness: From any malicious $\tilde{\mathcal{P}}$ we can extract an (α_n, R) -valid path except with ϵ prob.

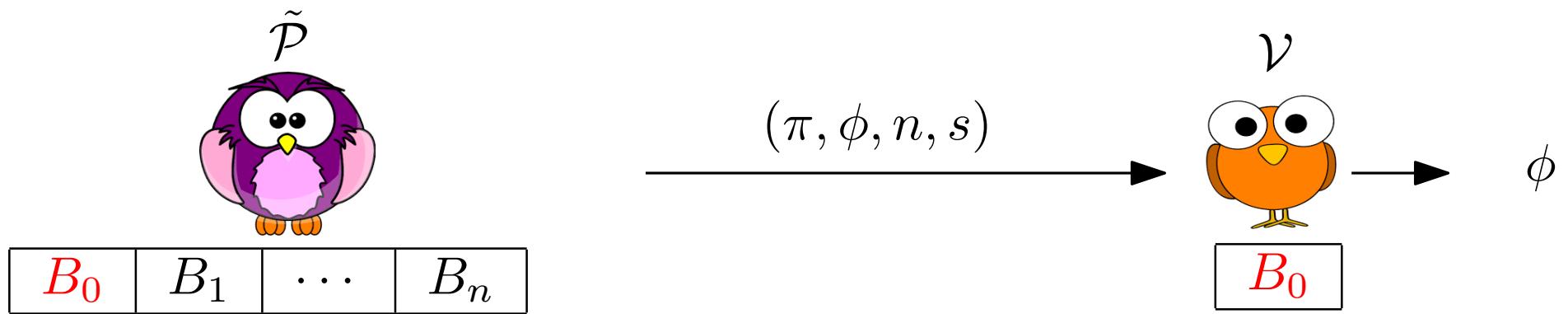
$$\alpha_n = 1 - \left(\log_c \left(\frac{\ell-1}{n+\ell} \right) \right)^{-1}$$



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Almost best possible guarantee: Meaningful for blockchains where it is moderately hard to generate a satisfying input for R

Thank you

