

# MuxProofs

Succinct Arguments for Machine Computation  
from Vector Lookups

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# Goal: Proofs for Machine Computation

Zero-knowledge, succinct proofs of knowledge of correct output of a program executed on specified input

## Program Description

```
Prog:  
  LOAD r1 0  
  MOV r1 r2  
  ADD r2 1  
  JMPIF r2 rx  
  Prog Instruction 5  
  Prog Instruction 6  
  ...  
  Prog Instruction 14  
  Prog Instruction 15  
  Prog Instruction 16
```

## Fixed Instruction Set

```
  ADD  
  SUB  
  MOV  
  MUL  
  LOAD  
  JMPIF  
  Instruction 7  
  Instruction 8
```

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```

## Program Execution Trace

```
Prog(input):  
  LOAD r1 0 (r1 <- input)  
  MOV r1 r2  
  ADD r2 1  
  JMPIF r2 rx  
  ...  
  Executed Instruction 30  
  Executed Instruction 31  
  Executed Instruction 32  
Returns output
```

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Statement  $x = (\text{Prog}, \text{input}, \text{output})$

Witness  $w = \text{Execution trace}$

$(x, w) \in R$  iff

“Prog executed on input results in output”

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Zero-knowledge: Verifier does not learn anything about w

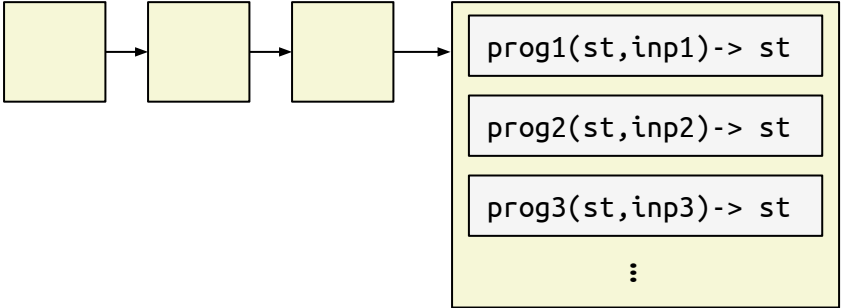
Proof of knowledge (soundness): Prover must know  $(x,w) \in R$  to convince verifier

Succinctness: Proof/verification is small w.r.t. program, instruction set, and execution trace

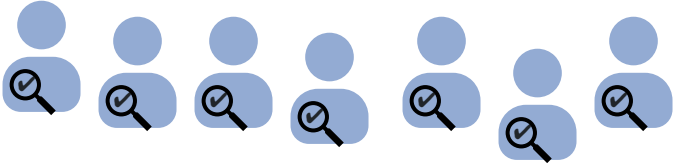
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# Application: Scalability for Blockchain Smart Contracts



prog1, inp1,  
prog2, inp2,  
prog3, inp3,  
...



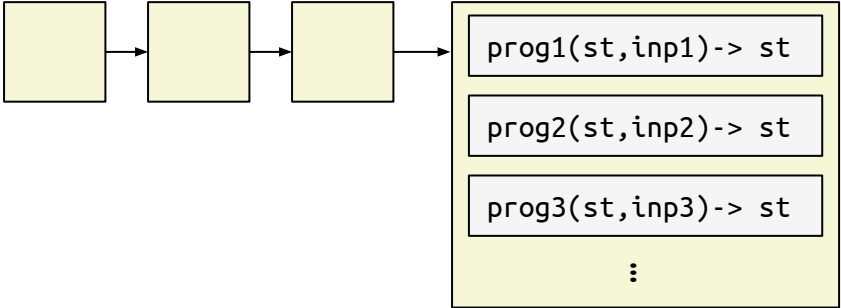
Validators

Blockchain is updated by applying smart contracts defined over supported instruction set, e.g., Ethereum Virtual Machine (EVM)

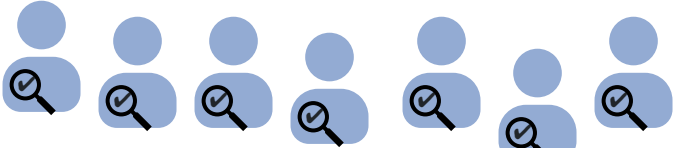
Contract data is dispersed across validators

Validators recompute contracts to ensure state has been updated correctly

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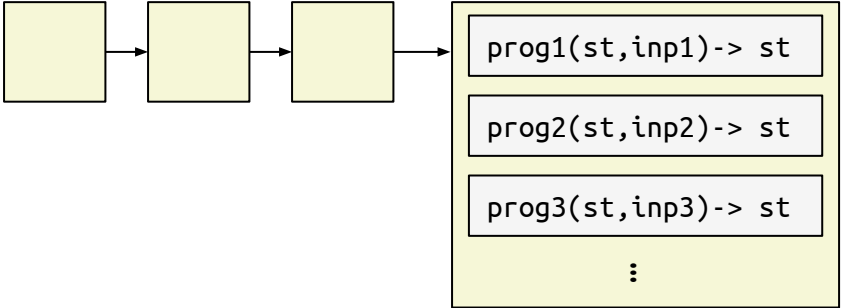
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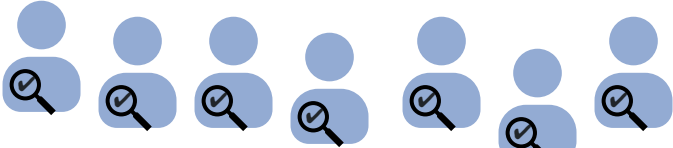
Validators recompute contracts to ensure state has been updated correctly

Succinct proofs of machine computation avoid redundancy in recomputation of contracts!

# Application: Privacy for Blockchain Smart Contracts



prog1, inp1,  
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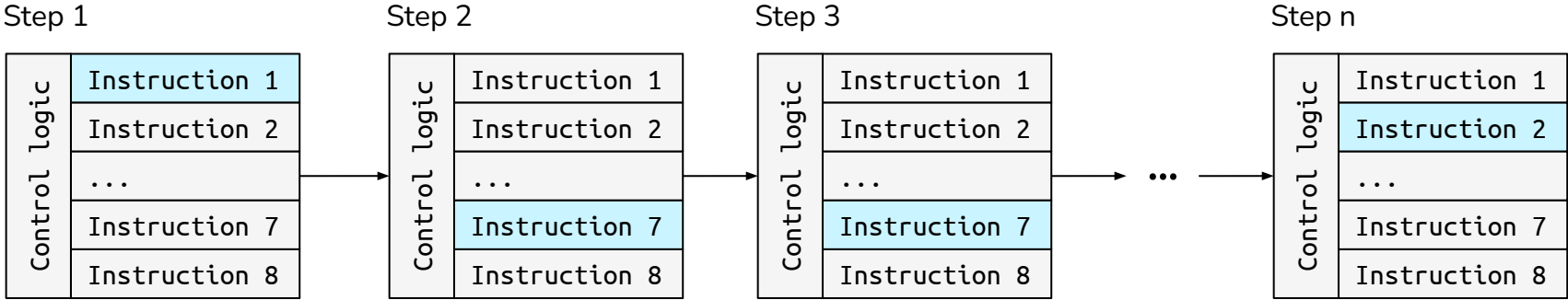
Zero-knowledge proofs of machine computation can hide contract and inputs allowing for privacy-preserving apps!



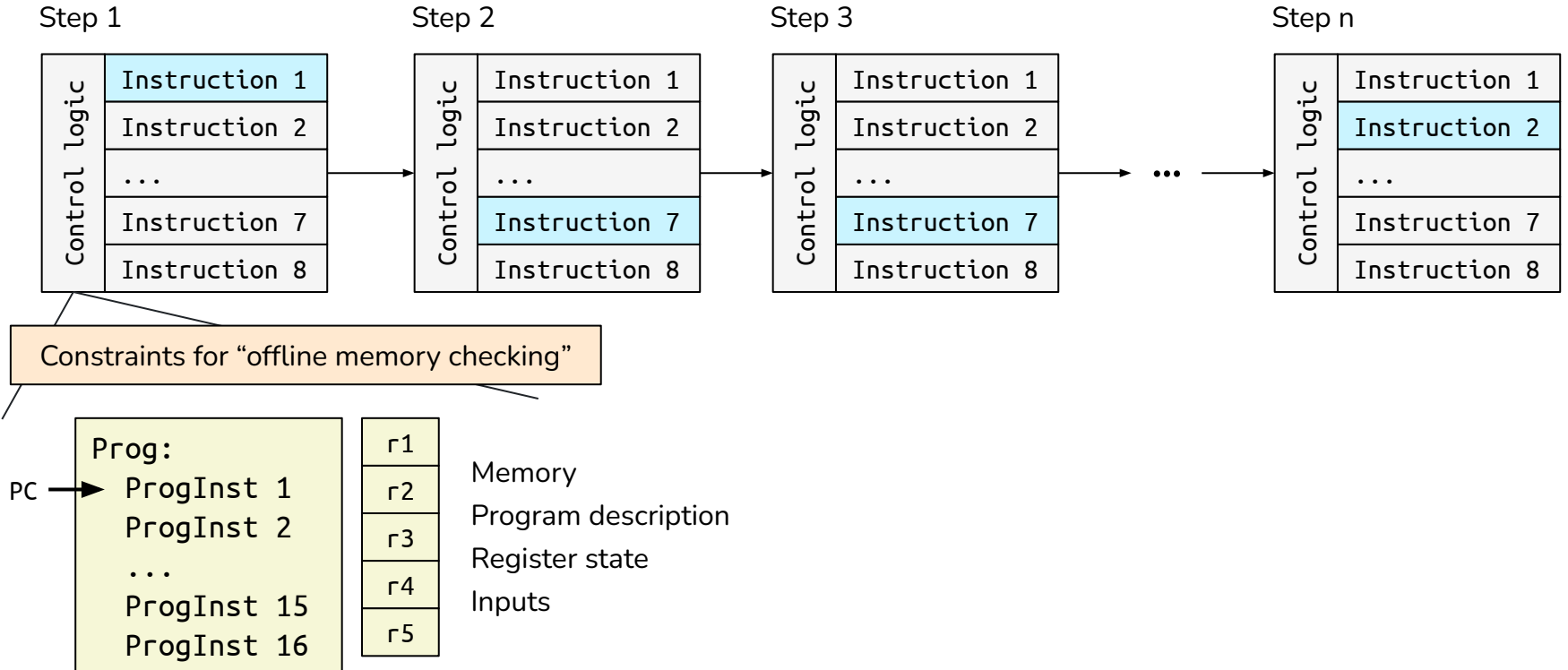
# This talk: MuxProofs

- New approach to proofs for machine execution
  - Prover costs scale with the cost of executed instructions
  - Avoids proof recursion
- New core cryptographic primitive: Succinct vector lookups
- Strategy:
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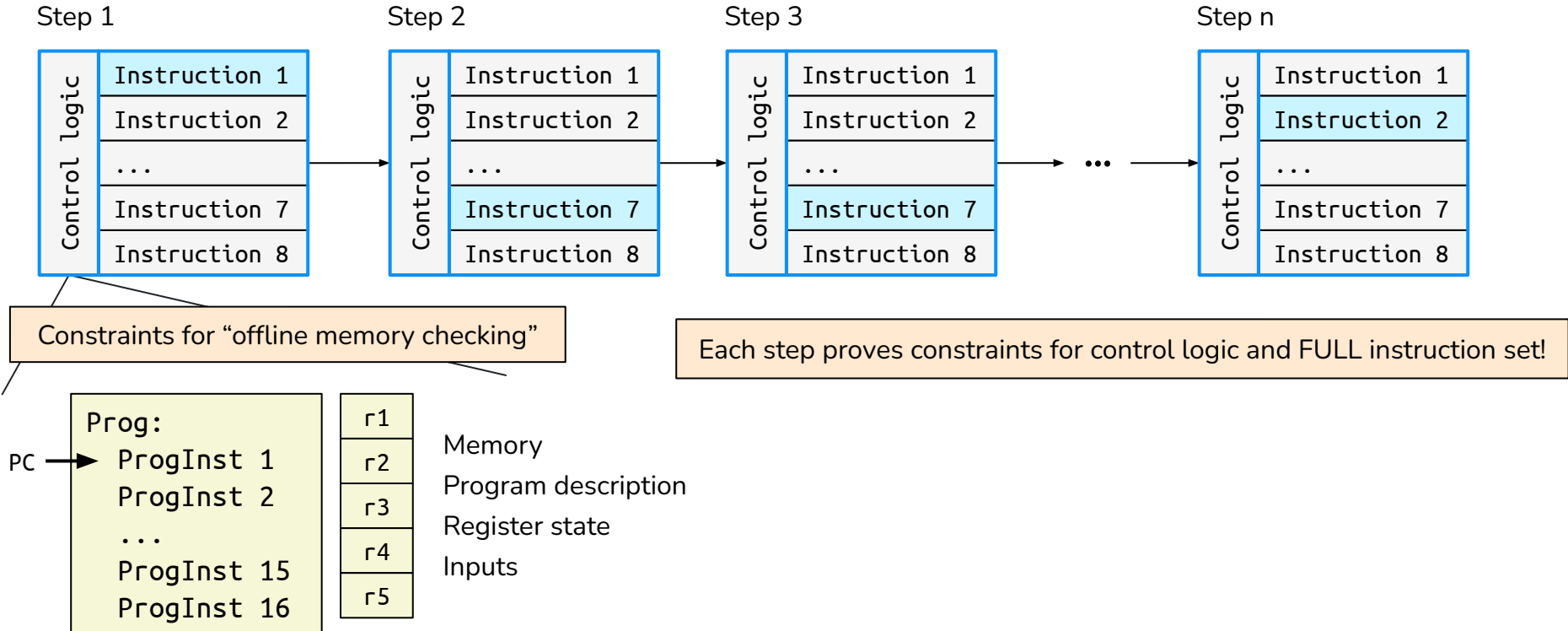
# Prior work: Universal instruction set constraints



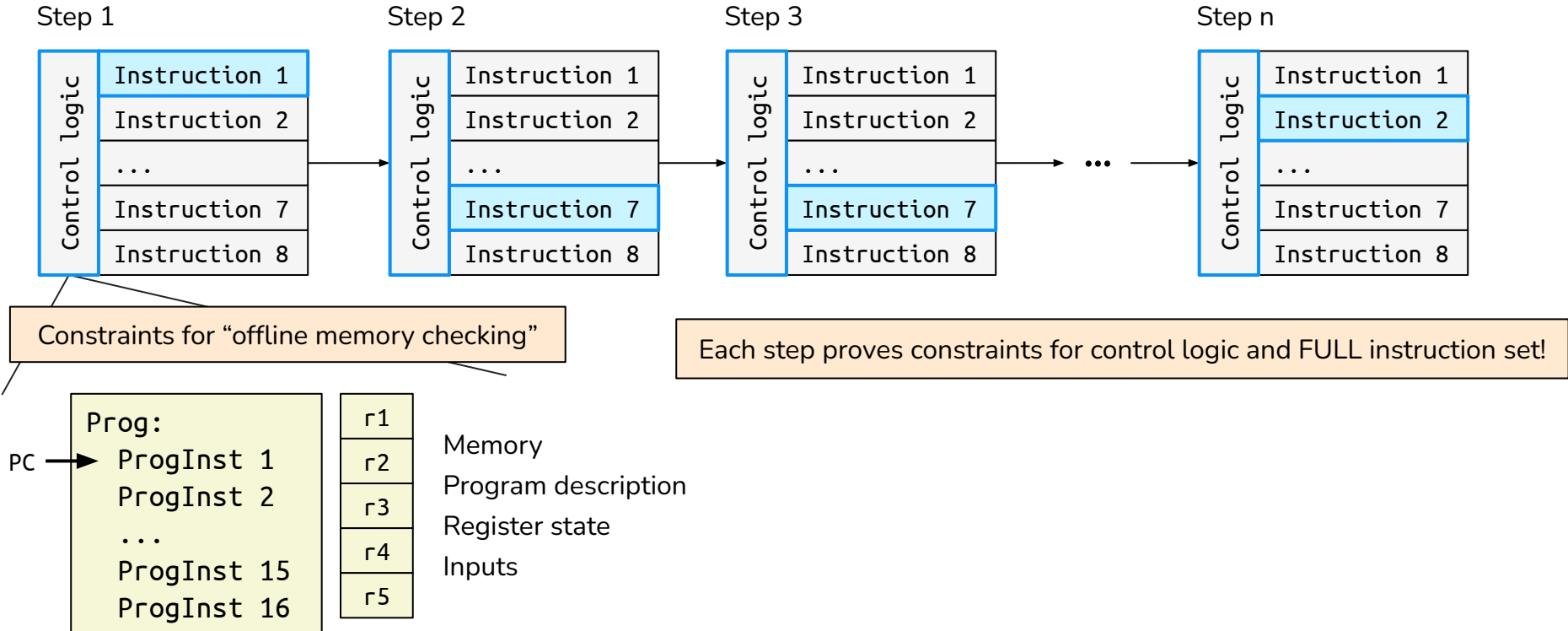
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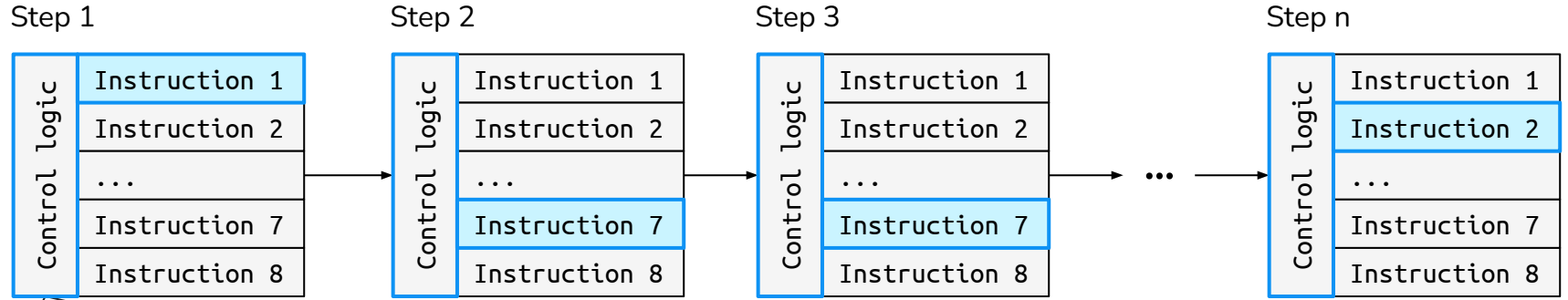
# Prior work: Universal instruction set constraints



# Goal: Executed instruction constraints

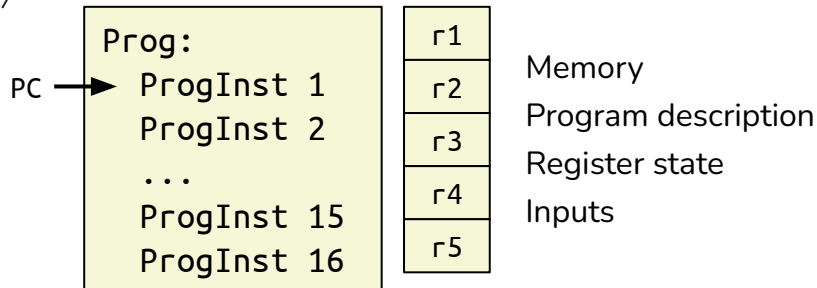


# Goal: Executed instruction constraints



Constraints for “offline memory checking”

Each step proves constraints for control logic and FULL instruction set!

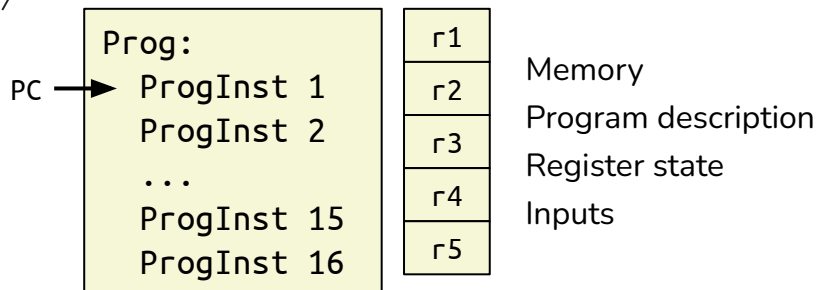
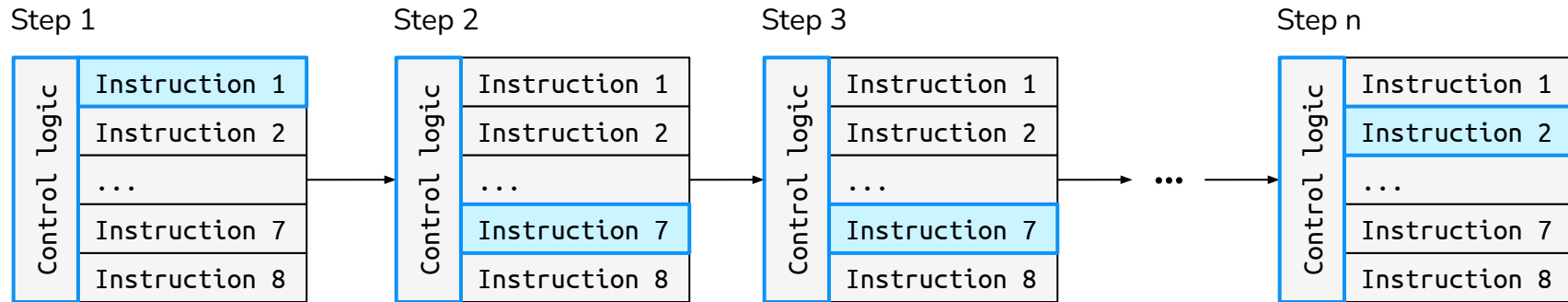


Hard to prove only constraints of executed instructions when:

- executed instructions depend on input
- want to provide zero-knowledge of executed instructions

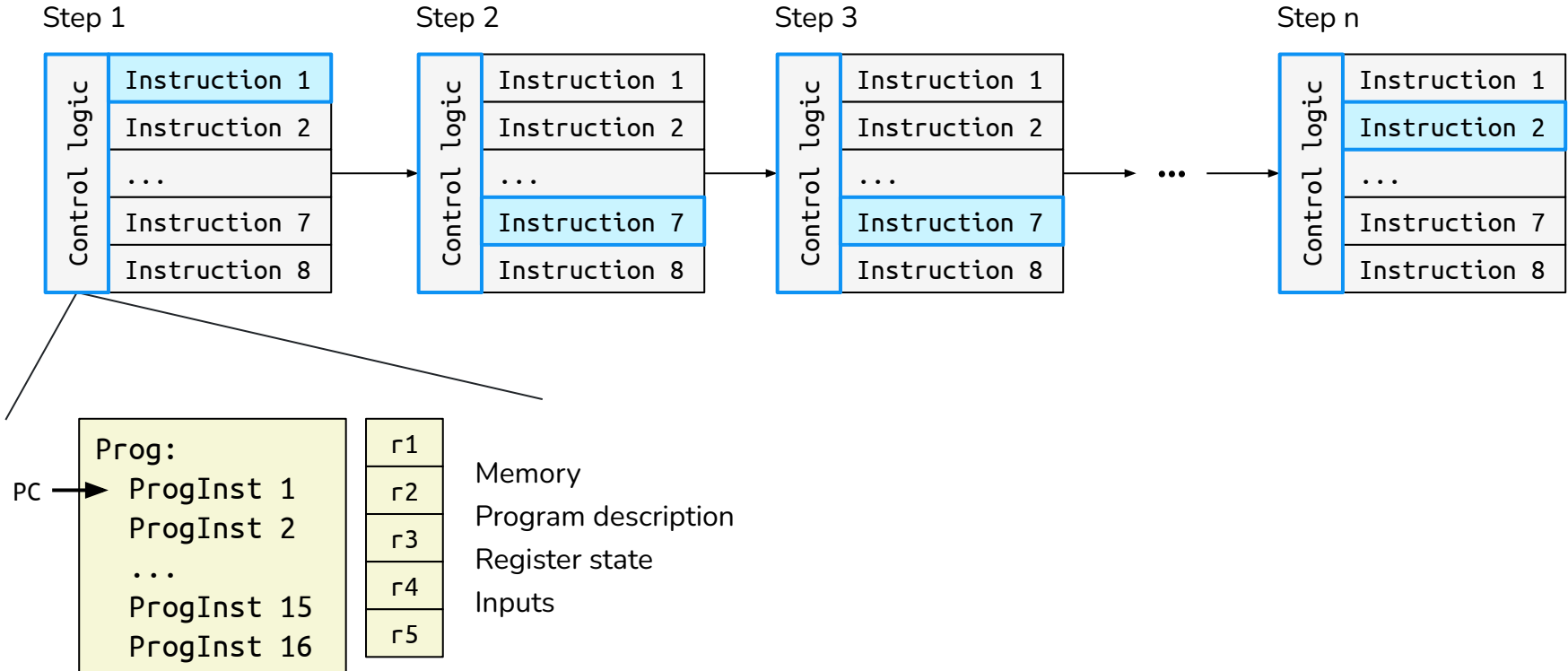
# Other work: Executed instruction constraints w/ recursion

SuperNova [Kothapalli & Setty] achieve proving costs on the order of executed constraints using proof recursion (folding)



# This work: Executed instruction constraints **w/o recursion**

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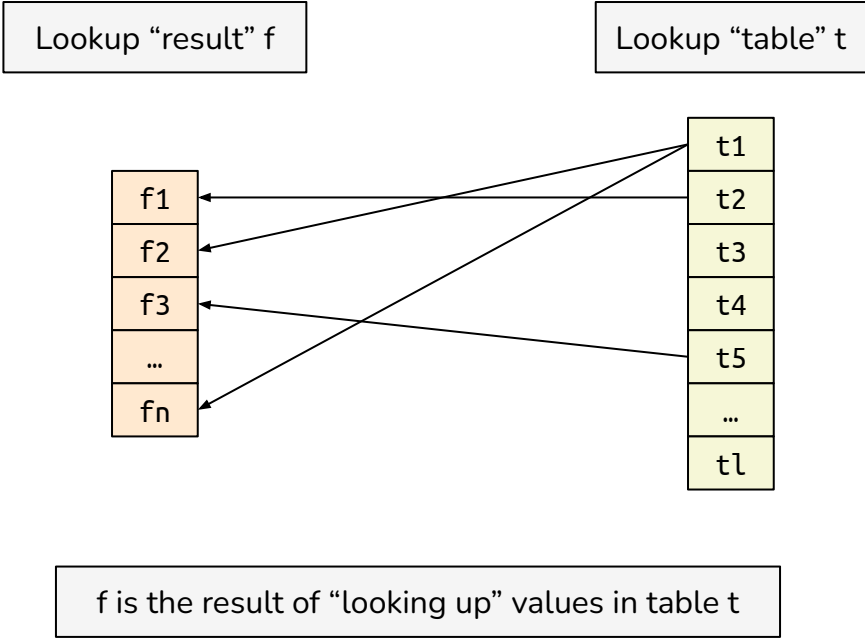




# This talk: MuxProofs

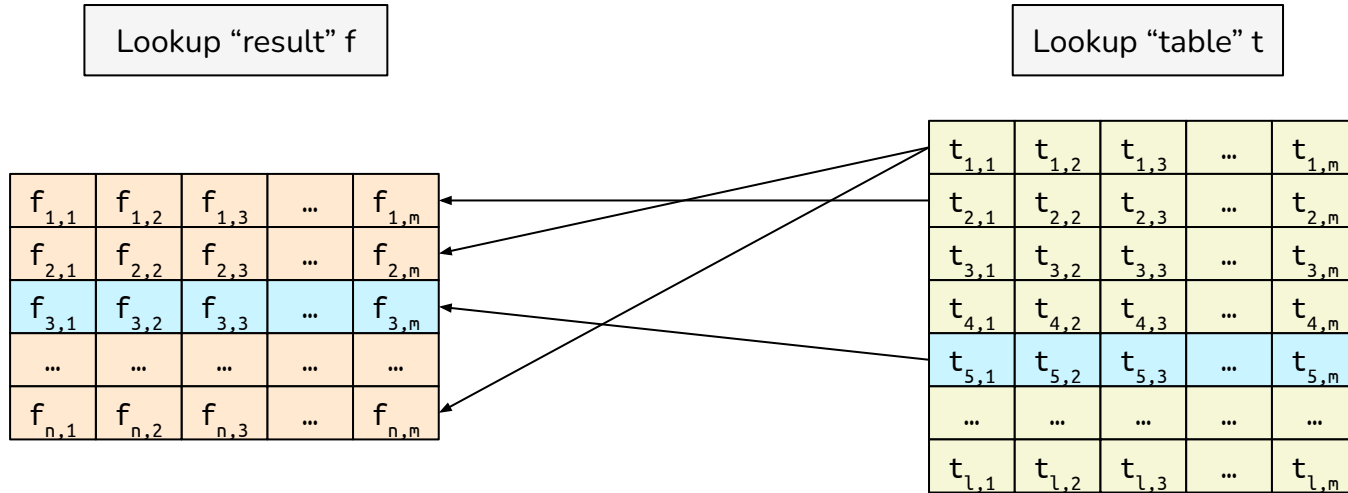
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# Key cryptographic primitive: Lookup arguments



$$\{f_i\}_{i \in [1, n]} \subseteq \{t_k\}_{k \in [1, l]}$$

# New cryptographic primitive: Succinct vector lookups

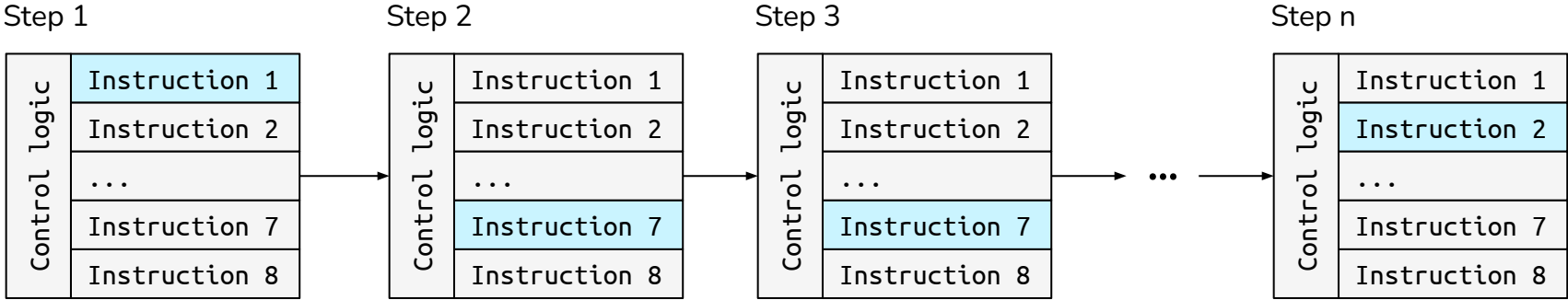


$f$  is the result of "looking up" **vectors** in table  $t$

$$\{[f_{i,1}, \dots, f_{i,m}]\}_{i \in [1, n]} \subseteq \{[t_{k,1}, \dots, t_{k,m}]\}_{k \in [1, l]}$$

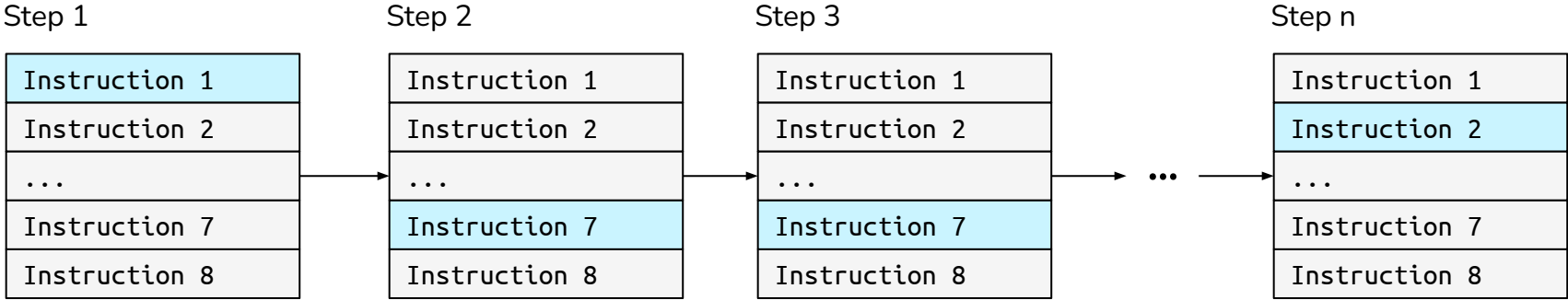
# Strategy: Vector lookups for machine computation

Detour: Preprocessing SNARKs based on polynomial encodings of vectors



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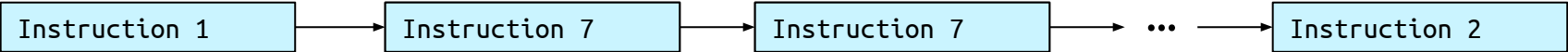
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Step 1

Step 2

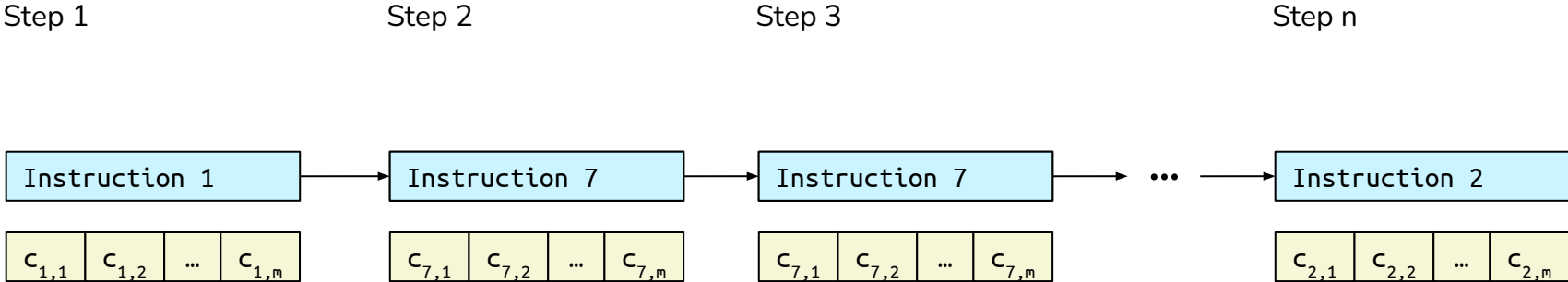
Step 3

Step n



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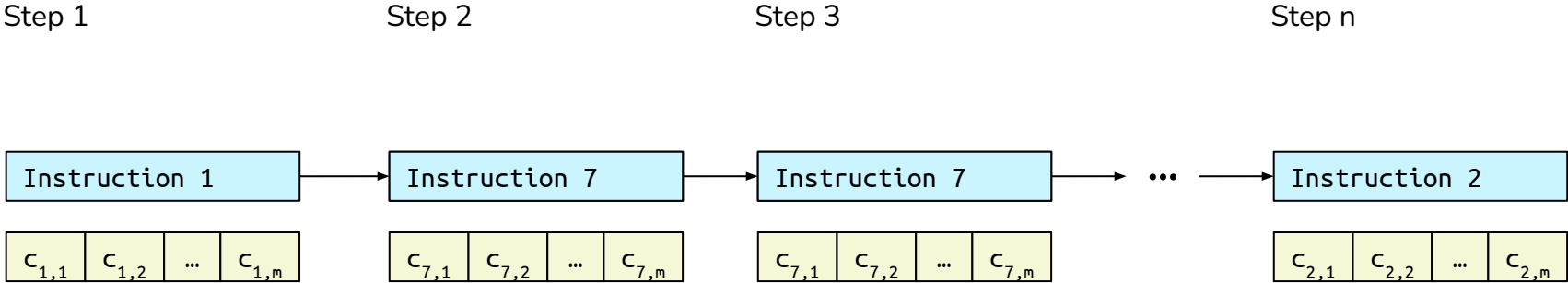


Constraints for an instruction are represented as a vector of field elements

Verifier preprocesses full vector of constraints as polynomial commitment for use in SNARK

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Detour: Preprocessing SNARKs based on polynomial encodings of vectors



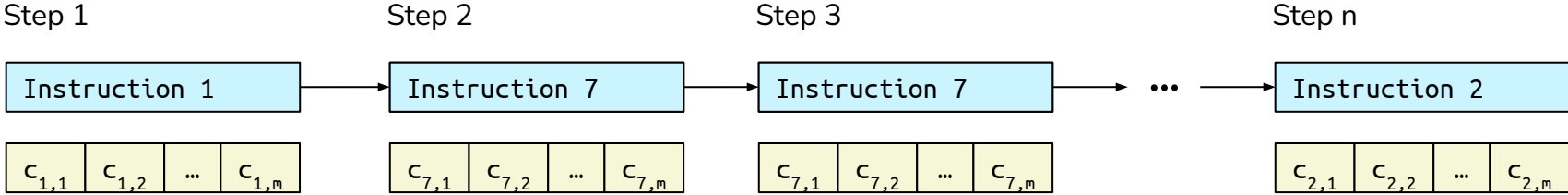
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Verifier cannot preprocess in machine computation setting since executed instruction order is not known ahead of time!



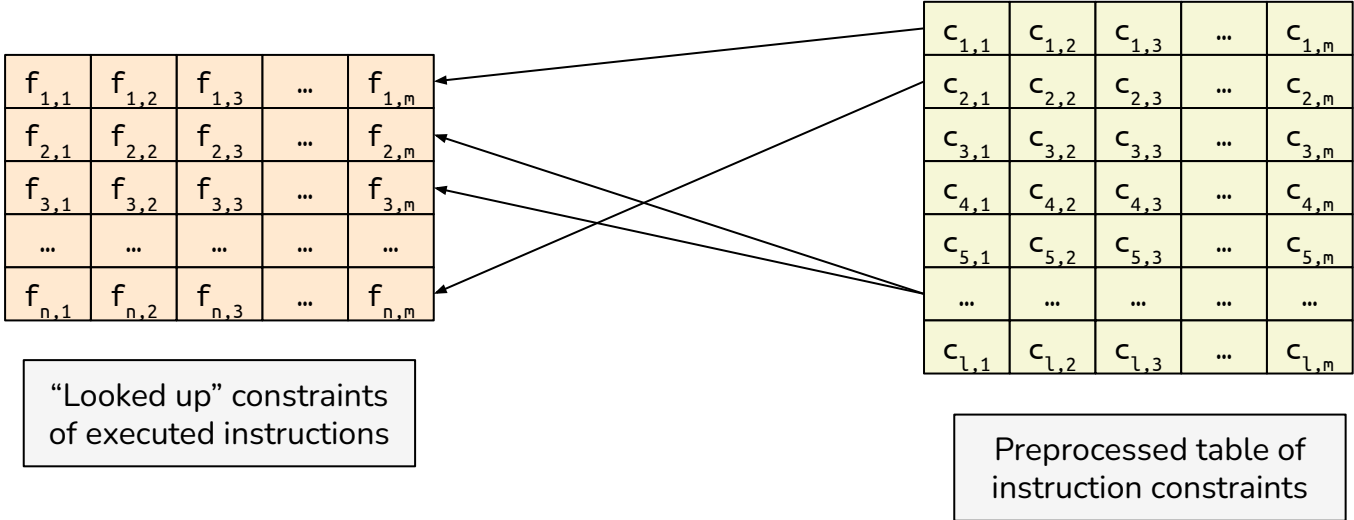
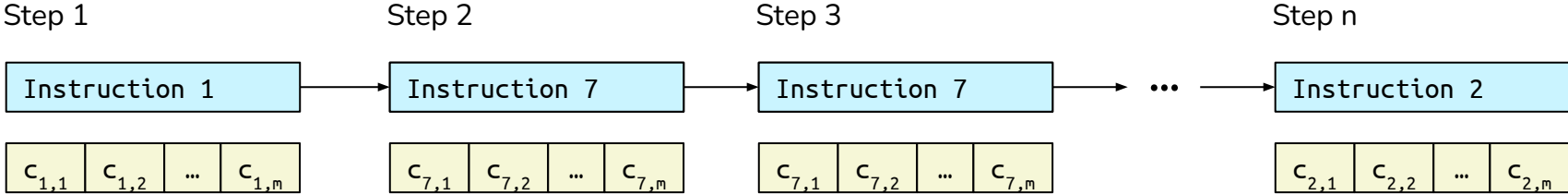
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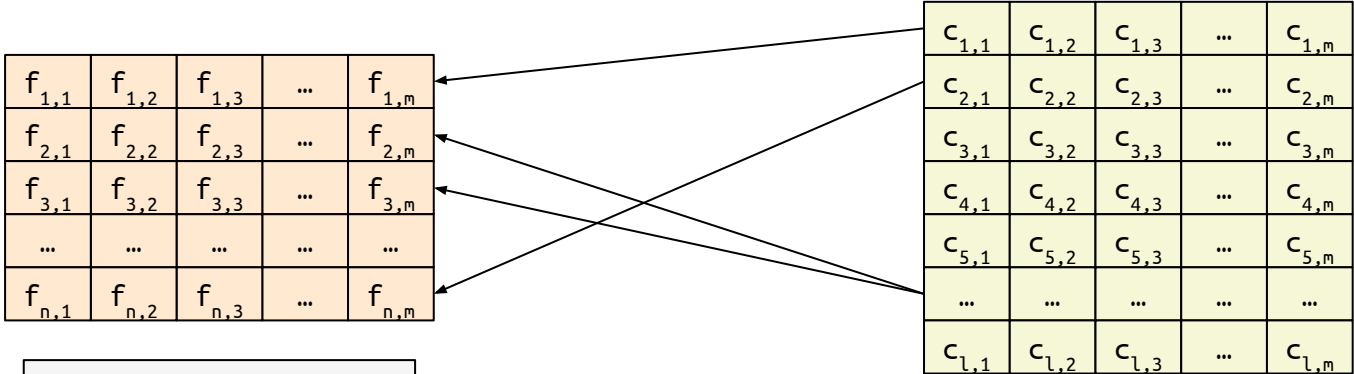
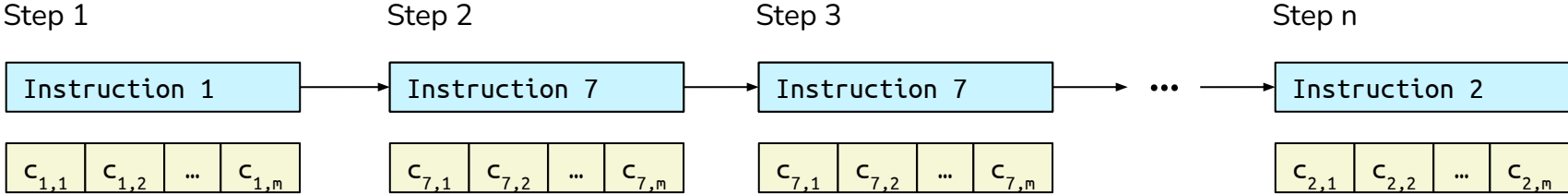
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...	...	...	...	...
$c_{l,1}$	$c_{l,2}$	$c_{l,3}$	...	$c_{l,m}$

Preprocessed table of instruction constraints

# Strategy: Vector lookups for machine computation



# Strategy: Vector lookups for machine computation



“Looked up” constraints of executed instructions

Preprocessed table of instruction constraints

Run SNARK on “f” as if it was a preprocessed vector of constraints

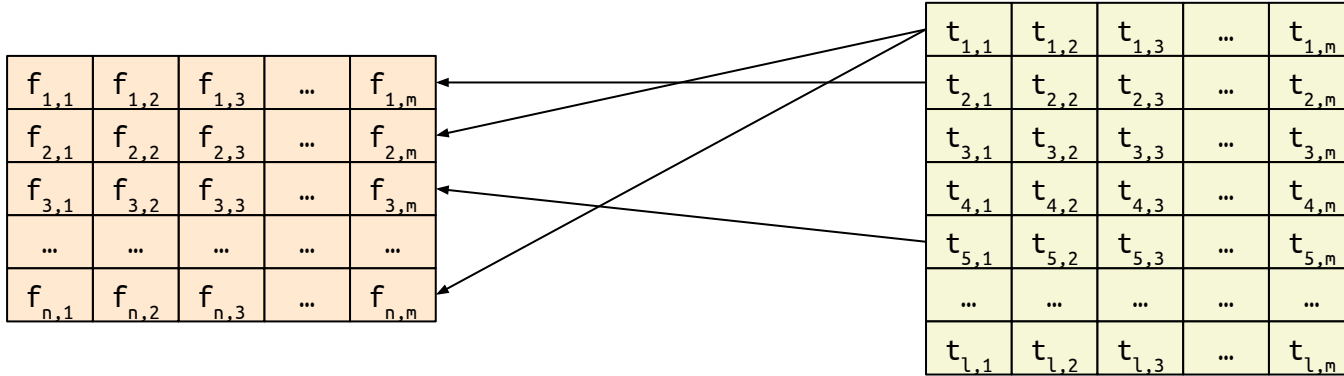
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# Succinct vector lookups via encodings over cosets

Important: Succinctness in “m” where “m” is the constraint length for an instruction.

- RISC-V: 32-bit instructions have  $m \approx 128$
- EVM: Some expensive instructions like SHA3 hash or ECDSA verification ( $m \approx 1.5$  mil)



$$\{[f_{i,1}, \dots, f_{i,m}]\}_{i \in [1, n]} \subseteq \{[t_{k,1}, \dots, t_{k,m}]\}_{k \in [1, l]}$$

# Prelims: Polynomial encoding of fields with cosets

$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$
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...	...	...	...	...
$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$

$f$ :

$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$
1	$\omega$	$\omega^2$	...	$\omega^{m-1}$	$\omega^m$	...								...	$\omega^{mn-1}$

Take subgroup of  $\Omega \subseteq F^*$  of size  $|\Omega| = mn$  generated by  $\Omega = \langle \omega \rangle$

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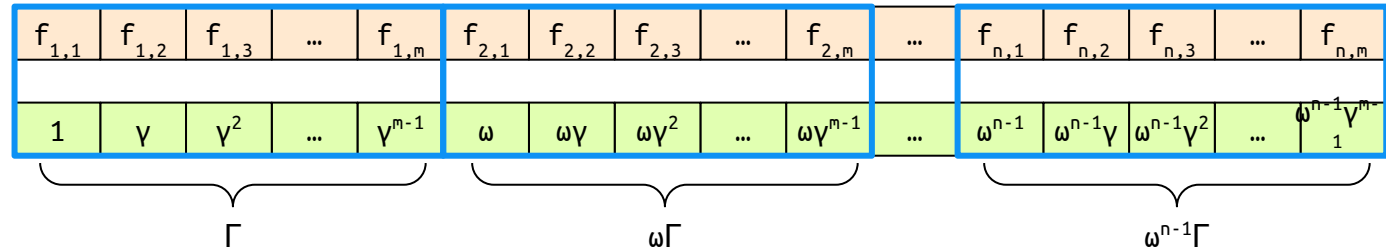
Does not provide algebraic structure to manipulate vectors

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$\gamma = \omega^n$



$n$  cosets of  $\Gamma$  in  $\Omega$ :

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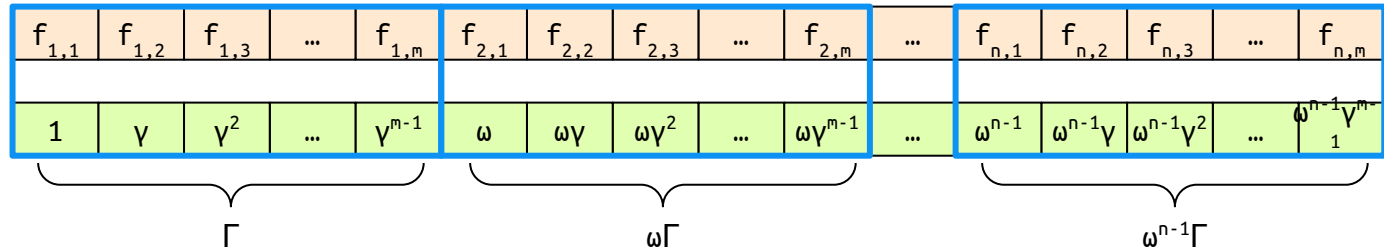
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Need cosets of size  $m$ ,  $nm$ , and  $lm$  fit to instruction set size and execution length!

$f$ :

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$n$  cosets of  $\Gamma$  in  $\Omega$ :

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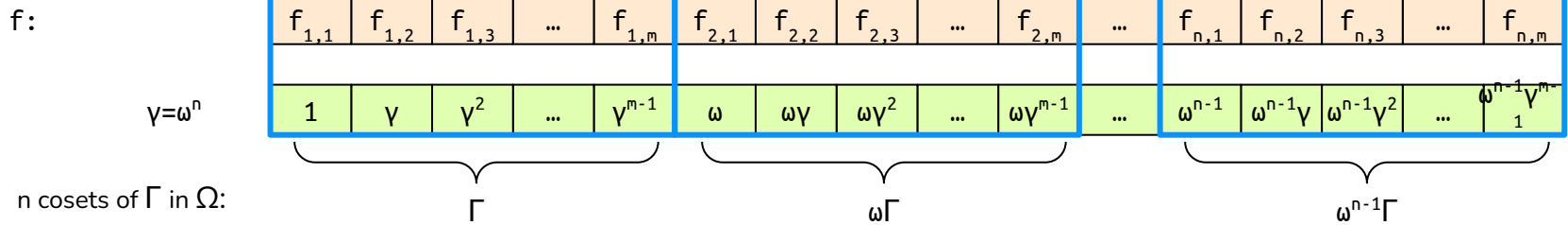
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Need cosets of size  $m$ ,  $nm$ , and  $lm$  fit to instruction set size and execution length!

Not a problem! Popular “FFT-friendly” pairing curves like BLS12-381 have  $F^*$  with subgroups of all powers-of-2 up to  $2^{32}$



Take subgroup of  $\Omega \subseteq F^*$  of size  $|\Omega| = mn$  generated by  $\Omega = \langle \omega \rangle$

Take subgroup of  $\Gamma \subseteq \Omega$  of size  $|\Gamma| = m$  generated by  $\Gamma = \langle \gamma \rangle$

# Construction preview: CosetLookup

Thm [Haböck '22]: If “f” and “t” are sequences of vectors the lookup relation holds

$$\{[f_{i,1}, \dots, f_{i,m}]\}_{i \in [1,n]} \subseteq \{[t_{k,1}, \dots, t_{k,m}]\}_{k \in [1,l]}$$

w.h.p. if there exists vector “c” that satisfies the following for random  $\alpha, \beta$  (selected after c)

$$\sum_{i \in [1,n]} \frac{1}{(\alpha + \sum_{j \in [1,m]} \beta^{j-1} \cdot f_{i,j})} = \sum_{k \in [1,l]} \frac{c_k}{(\alpha + \sum_{j \in [1,m]} \beta^{j-1} \cdot t_{k,j})}$$

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# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$

f:

$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$
1	$\gamma$	$\gamma^2$	...	$\gamma^{m-1}$	$\omega$	$\omega\gamma$	$\omega\gamma^2$	...	$\omega\gamma^{m-1}$	...	$\omega^{n-1}$	$\omega^{n-1}\gamma$	$\omega^{n-1}\gamma^2$	...	$\omega^{n-1}\gamma^{m-1}$

$\gamma = \omega^n$

n cosets of  $\Gamma$  in  $\Omega$ :

$\Gamma$

$\omega\Gamma$

$\omega^{n-1}\Gamma$

# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$

f:

	$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$
$\gamma = \omega^n$	1	$\gamma$	$\gamma^2$	...	$\gamma^{m-1}$	$\omega$	$\omega\gamma$	$\omega\gamma^2$	...	$\omega\gamma^{m-1}$	...	$\omega^{n-1}$	$\omega^{n-1}\gamma$	$\omega^{n-1}\gamma^2$	...	$\omega^{n-1}\gamma^{m-1}$
n cosets of $\Gamma$ in $\Omega$ :	$\Gamma$					$\omega\Gamma$					$\omega^{n-1}\Gamma$					
I:				...					...		...				...	

# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$

f:	$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$	
$\gamma = \omega^n$	1	$\gamma$	$\gamma^2$	...	$\gamma^{m-1}$	$\omega$	$\omega\gamma$	$\omega\gamma^2$	...	$\omega\gamma^{m-1}$	...	$\omega^{n-1}$	$\omega^{n-1}\gamma$	$\omega^{n-1}\gamma^2$	...	$\omega^{n-1}\gamma^{m-1}$	1
n cosets of $\Gamma$ in $\Omega$ :	$\Gamma$					$\omega\Gamma$					$\omega^{n-1}\Gamma$						
I:	1			...					...		...				...		

$$I(1) = 1$$

# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$

f:	$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$
$\gamma = \omega^n$	1	$\gamma$	$\gamma^2$	...	$\gamma^{m-1}$	$\omega$	$\omega\gamma$	$\omega\gamma^2$	...	$\omega\gamma^{m-1}$	...	$\omega^{n-1}$	$\omega^{n-1}\gamma$	$\omega^{n-1}\gamma^2$	...	$\omega^{n-1}\gamma^{m-1}$
n cosets of $\Gamma$ in $\Omega$ :	$\Gamma$					$\omega\Gamma$					$\omega^{n-1}\Gamma$					
I:	1	$\beta$	$\beta^2$	...	$\beta^{m-1}$				...		...				...	

$$I(1) = 1$$

$$(I(\gamma X) - \beta \cdot I(X))(X - \gamma^{m-1}) = 0 \quad \text{over } \Gamma$$



# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$

f:	$f_{1,1}$	$f_{1,2}$	$f_{1,3}$	...	$f_{1,m}$	$f_{2,1}$	$f_{2,2}$	$f_{2,3}$	...	$f_{2,m}$	...	$f_{n,1}$	$f_{n,2}$	$f_{n,3}$	...	$f_{n,m}$
$\gamma = \omega^n$	1	$\gamma$	$\gamma^2$	...	$\gamma^{m-1}$	$\omega$	$\omega\gamma$	$\omega\gamma^2$	...	$\omega\gamma^{m-1}$	...	$\omega^{n-1}$	$\omega^{n-1}\gamma$	$\omega^{n-1}\gamma^2$	...	$\omega^{n-1}\gamma^{m-1}$
n cosets of $\Gamma$ in $\Omega$ :	$\Gamma$					$\omega\Gamma$					$\omega^{n-1}\Gamma$					
I:	1	$\beta$	$\beta^2$	...	$\beta^{m-1}$	1	$\beta$	$\beta^2$	...	$\beta^{m-1}$	...	1	$\beta$	$\beta^2$	...	$\beta^{m-1}$

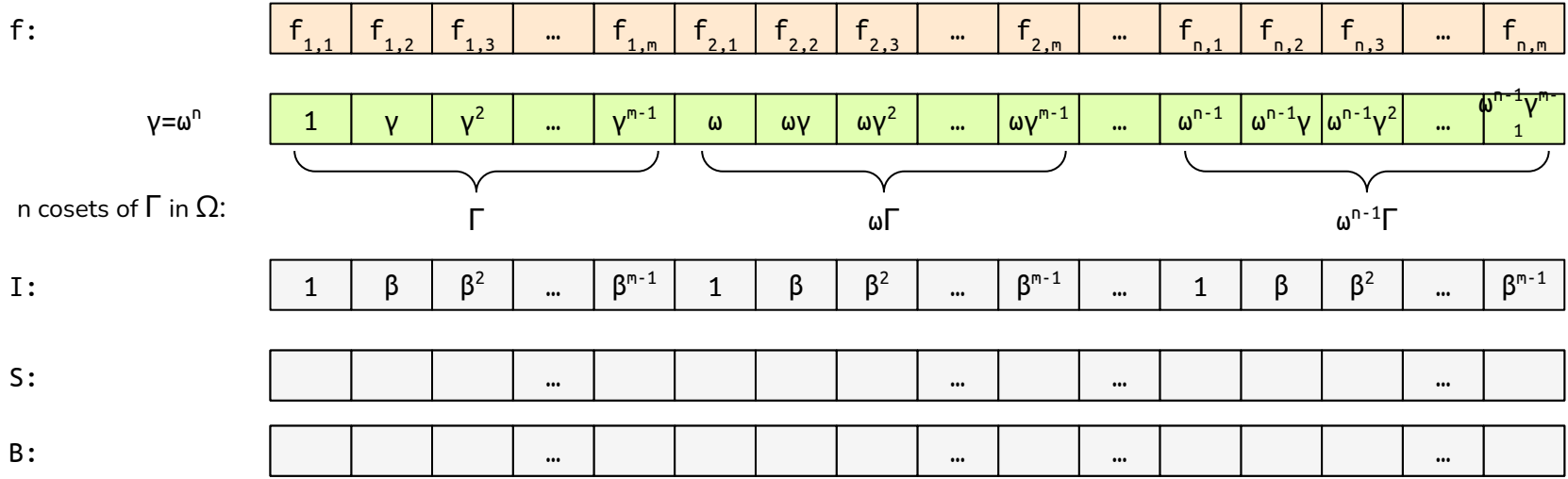
$$I(1) = 1$$

$$(I(\gamma X) - \beta \cdot I(X))(X - \gamma^{m-1}) = 0 \quad \text{over } \Gamma$$

$$(I(\omega X) - I(X))(Z[\omega^{n-1}\Gamma](X) - \gamma^{m-1}) = 0 \quad \text{over } \Omega$$

# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$



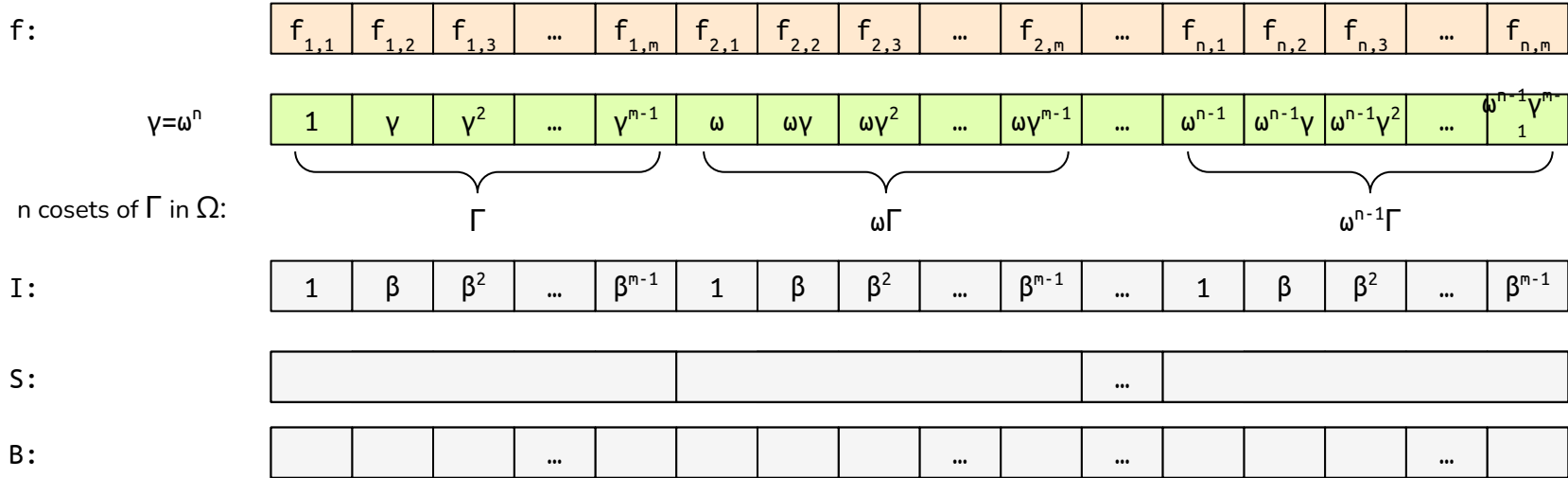
$I(1) = 1$

$(I(\gamma X) - \beta \cdot I(X))(X - \gamma^{m-1}) = 0 \quad \text{over } \Gamma$

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# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$



$$I(1) = 1$$

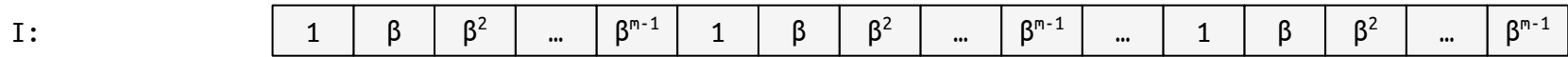
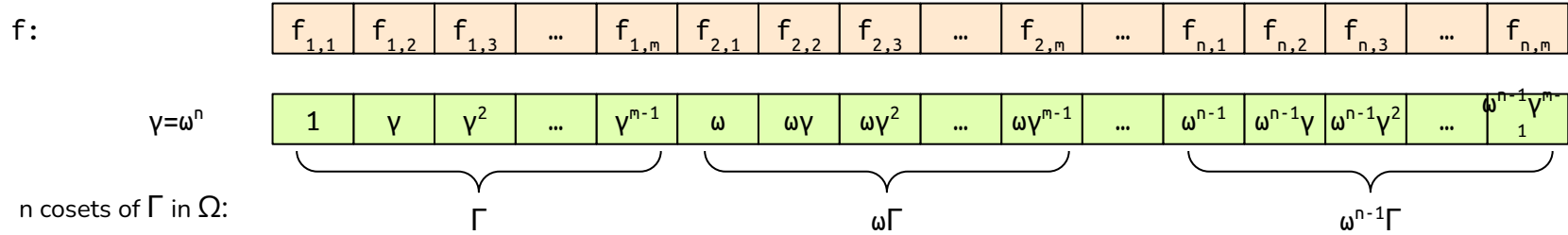
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$$S(\gamma X) = S(X) \quad \text{over } \Omega$$

# Construction preview: CosetLookup

$$\sum_{i \in [1, n]} \frac{1}{(a + \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j})}$$



$I(1) = 1$

$(I(\gamma X) - \beta \cdot I(X))(X - \gamma^{m-1}) = 0$  over  $\Gamma$

$(I(\omega X) - I(X))(Z[\omega^{n-1}\Gamma](X) - \gamma^{m-1}) = 0$  over  $\Omega$

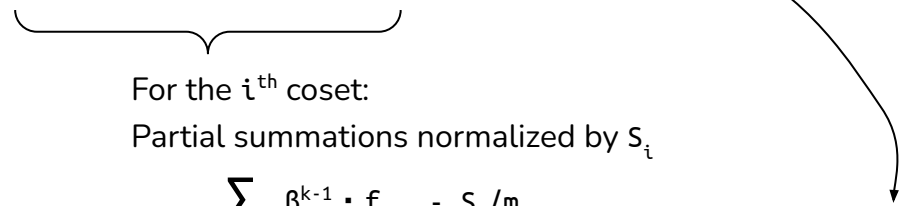
$S(\gamma X) = S(X)$  over  $\Omega$

$B(\gamma X) = B(X) + I(\gamma X) \cdot B(\gamma X) - S(X)/m$  over  $\Omega$

For the  $i^{\text{th}}$  coset:  
 Partial summations normalized by  $S_i$

$$\sum_{k \in [1, j]} \beta^{k-1} \cdot f_{i,k} - S_i/m$$

$$S_i = \sum_{j \in [1, m]} \beta^{j-1} \cdot f_{i,j}$$



# This talk: MuxProofs

- New approach to proofs for machine execution
  - Prover costs scale with the cost of executed instructions
  - Avoids proof recursion
- New core cryptographic primitive: Succinct vector lookups
  - Univariate setting: Coset encoding
  - Multivariate setting: Boolean subcube encoding
- Strategy:
  - Machine computation from succinct vector lookups
  - Succinct vector lookups from polynomial encodings on structured evaluation subdomains

# Reserve Slides

# Disclaimer: Why avoid recursion?

Cons of recursion

No security proof in random oracle model

Pros of recursion

Small memory usage

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No security proof in random oracle model

Best encodings use cycles of elliptic curves

Recursive overhead at each step

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Modern recursion techniques (folding) don't "prove" all constraints, instead "folds" them together into something of smaller size

<

For proving speed, *probably* state-of-the-art recursive approaches [SuperNova] would outperform MuxProofs

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We use this observation in Mangrove [CRYPTO '24] where we apply recursion to general computation

# Example: Lookups used in proofs for machine computation

Decrease number of constraints in complex instructions with contained “lookups” like SHA2

SHA2 Arithmetic Circuit Constraints

>

SHA2 Arithmetic Circuit Constraints  
without lookups

+

SHA2 Lookup Constraints

Directly “lookup” instruction result if instruction has certain algebraic structure, e.g., ADD in RISC-V

$2^{32}$ -bit ADD  
Arithmetic Circuit Constraints

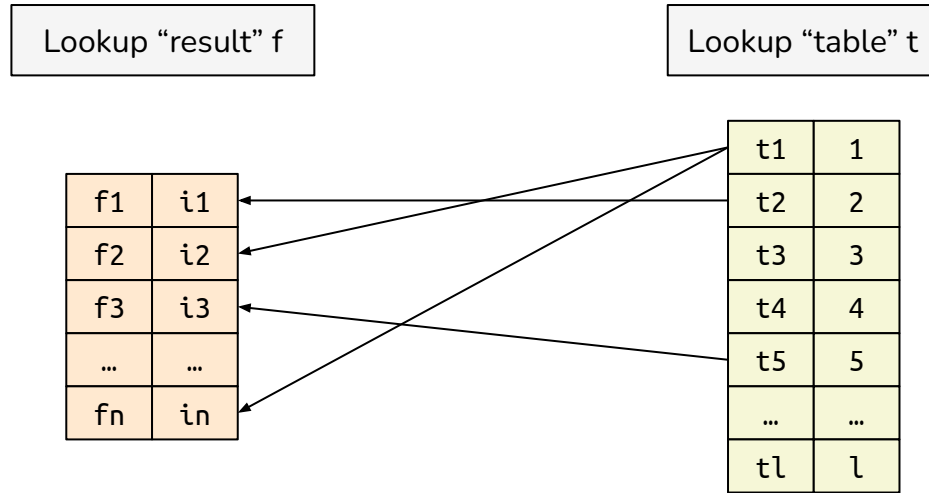
>

$2^{32}$ -bit ADD  
Lookup Constraints  
( $2^{64}$ -size table)

[Lasso, Jolt EUROCRYPT '24]

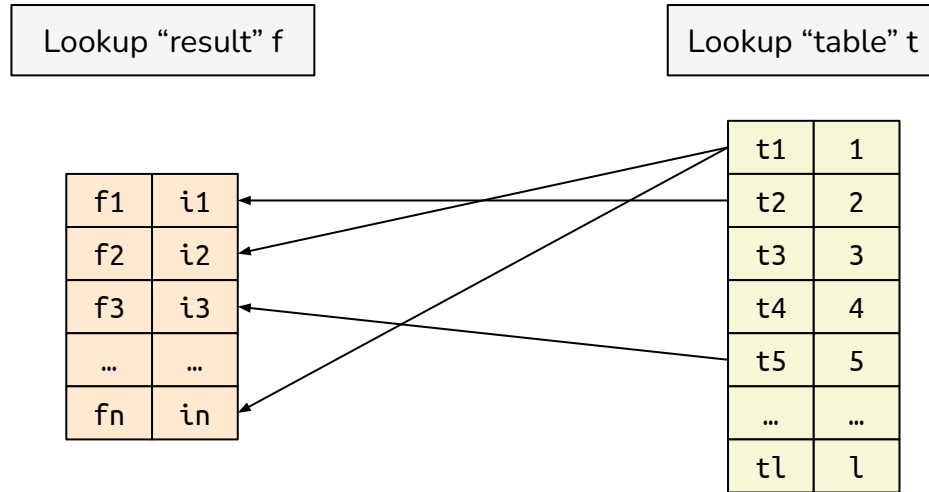
We use lookups differently! Our ideas are orthogonal and can likely be composed with other strategies

# Example: Vector lookup arguments specifying index



Vectors of length 2 can be used to enforce lookups for a specified index!

# Example: Vector lookup arguments specifying index



Vectors of length 2 can be used to enforce lookups for a specified index!

Can build vector lookups from element lookups [Plookup '20], but not succinct in vector length