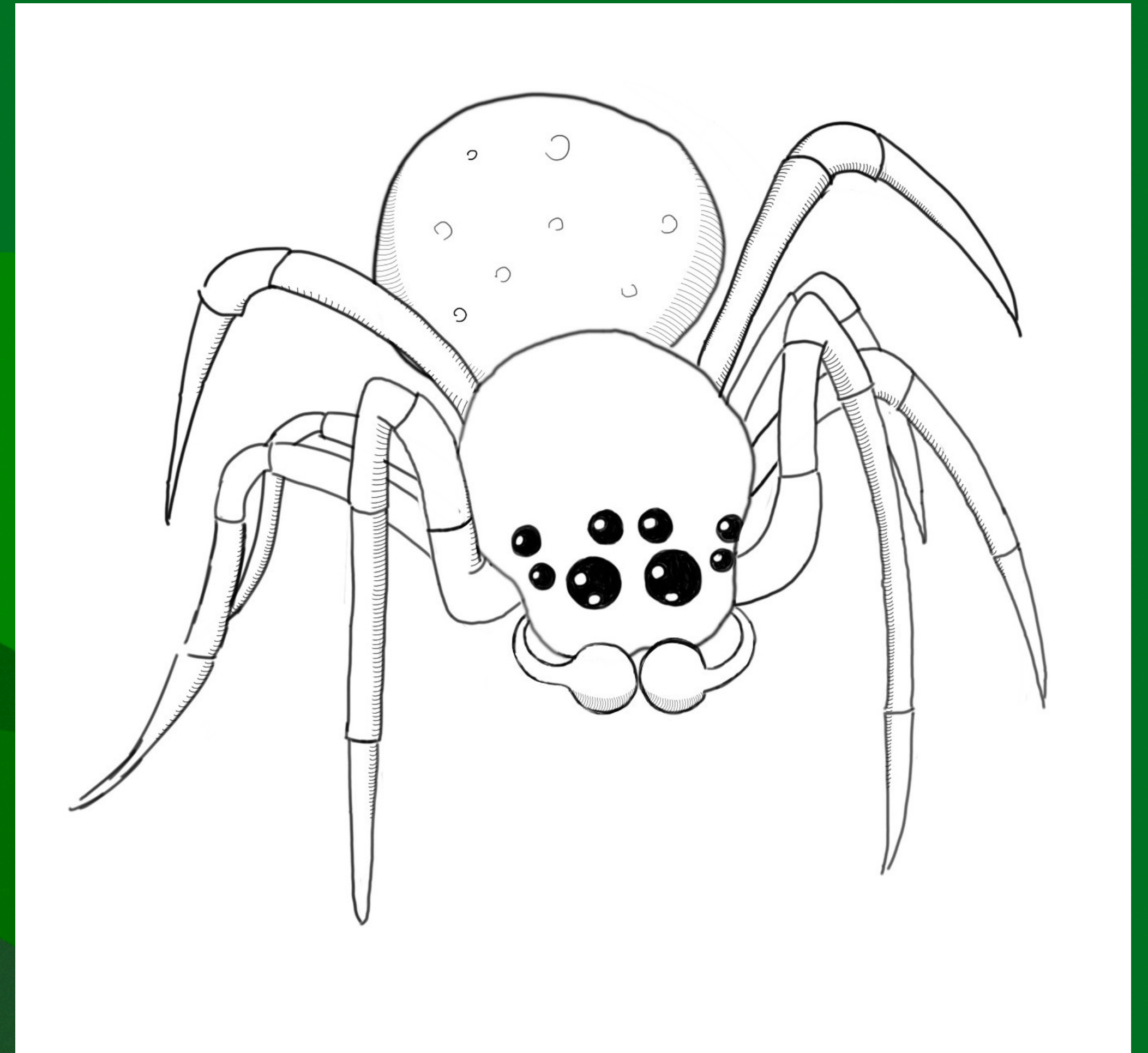


Orbweaver

Succinct Linear Functional Commitments from Lattices

Ben Fisch, Zeyu Liu, and **Psi Vesely**

Yale University



Lattice Orbweaver Spider by Jackie P (CC BY 4.0)

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Abstract linear map equation

$$\left(\sum_{i=1}^n f_i \cdot Y^{-i} \right) \cdot \left(\sum_{i=1}^n x_i \cdot Y^i \right) \equiv \langle \mathbf{f}, \mathbf{x} \rangle + \sum_{\substack{i=-n+1, \\ i \neq 0}}^{n-1} b_i \cdot Y^i \pmod{q}$$

Form used in [Gro10,LRY16,AC20]

Evaluation verification equation

\mathbf{f}, \mathbf{x} short

$$\left(\sum_{i=1}^n f_i \cdot Y^{-i} \right) \cdot \left(\sum_{i=1}^n x_i \cdot Y^i \right) \equiv \langle \mathbf{f}, \mathbf{x} \rangle + \sum_{\substack{i = -n+1, \\ i \neq 0}}^{n-1} b_i \cdot Y^i \pmod{q}$$

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$c_{\mathbf{f}}$

\cdot

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Form used in [Gro10,LRY16,AC20], translated to lattice setting using techniques from [ACLMT22]

Ring Vandermonde SIS (R-V-SIS) commitment

$$c := \sum_{i=1}^n x_i \cdot v^i \pmod{q}, \text{ where } v \stackrel{\$}{\leftarrow} R_q \text{ is public}$$

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- Ajtai's R-SIS commitment, with a Vandermonde key
- Similar to assumption used in PASS Sign. If we pick v instead from the primitive roots of unity binding reduces to Vandermonde R-SIS [HS15,LZA18,BSS22]

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$$c_{\mathbf{f}} \cdot c_{\mathbf{x}} \equiv y$$

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$+$

$\langle \mathbf{a}, \boldsymbol{\pi} \rangle \pmod{q}$

(preprocessed)

$\boldsymbol{\pi}$ short

Form used in [Gro10,LRY16,AC20], translated to lattice setting using techniques from [ACLMT22]

Prover key

Generate short preimages \mathbf{u}_i for $i \in \{-n + 1, \dots, n - 1\} \setminus \{0\}$ such that

$$\langle \mathbf{a}, \mathbf{u}_i \rangle \equiv v^i \pmod{q}$$

Using a trapdoor public SIS matrix \mathbf{a} [MP12]

Computing the proof

- Given $\langle \mathbf{a}, \mathbf{u}_i \rangle \equiv v^i \pmod{q}$ except for $i = 0$
- Where b_i is the sum of cross terms corresponding to the coefficient of v^i compute

$$\pi := \sum_{\substack{i = -n + 1, \\ i \neq 0}}^{n-1} b_i \cdot \mathbf{u}_i \pmod{q}$$

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$$\langle \mathbf{a}, \pi \rangle \equiv \sum_{\substack{i = -n+1, \\ i \neq 0}}^{n-1} b_i \cdot v^i \pmod{q}$$

- \mathbf{f}, \mathbf{x} short $\implies b_i$ short, u_i short $\implies \pi$ short

Evaluation binding

$$\left(\sum_{i=1}^n f_i \cdot v^{-i} \right) \cdot \left(\sum_{i=1}^n x_i \cdot v^i \right) \equiv \langle \mathbf{f}, \mathbf{x} \rangle + \sum_{\substack{i=-n+1, \\ i \neq 0}}^{n-1} b_i \cdot v^i \pmod{q}$$

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$$c_{\mathbf{f}} \cdot c_{\mathbf{x}} \equiv y + \langle \mathbf{a}, \pi \rangle \pmod{q}$$

$$\langle \mathbf{a}, \pi - \pi' \rangle \equiv y' - y \pmod{q}$$

Evaluation binding

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- k-R-ISIS family of assumptions: can only generate short preimages for targets within a short linear span of the v^i or for random targets [ACLMT22]

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- k-R-ISIS family of assumptions: can only generate short preimages for targets within a short linear span of the v^i or for random targets [ACLMT22]
- $y' - y$ is short, while for $v \xleftarrow{\$} R_q$ all $v^i \pmod{q}$ will be long whp, as will the random targets

Multiple outputs

Can prove $\langle \mathbf{f}_i, \mathbf{x} \rangle = y_i$ for $i \in [t]$ with a single evaluation proof:

$$\langle \mathbf{a}, \pi \rangle \equiv c \cdot \sum_{i=1}^t h_i \cdot \text{ck}_{\mathbf{f}_i} - \sum_{i=1}^t h_i \cdot y_i \pmod{q}$$

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Key observation: the prover submits a separate knowledge proof π' for c from which we extract \mathbf{x} . It's thus unnecessary to extract the hypothetical π_i s.t.

$$\pi = \sum_{i=1}^t h_i \cdot \pi_i$$

Multiple outputs

Using extracted \mathbf{x} we get

$$\langle \mathbf{a}, \pi \rangle \equiv \sum_{i=1}^t h_i \cdot (\langle \mathbf{f}_i, \mathbf{x} \rangle - y_i) - \sum_{i=1}^t h_i \cdot \sum_{\substack{j = -n + 1, \\ j \neq 0}}^{n-1} b_{i,j} \cdot Y^i \pmod{q}$$

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$$p(h_1, \dots, h_t)$$

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- Can pick *exponential size* “exceptional set” \mathcal{H} over R_q for large q [LS18] and invoke Generalized Alon-Füredi Theorem [BCPS18]
- Better to perform ternary decomposition on \mathbf{f}, \mathbf{x} and batch verification

Proof and SRS sizes for $\mathbb{Z}_{2^{32}}$

$\log_2(x)$	18	22	26	30
$ c $ (B)	293	347	422	505
total proof size (KiB)	845	1,081	1,315	1,571
verifier key (MiB)	12	17	23	30
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- Evaluation binding only version (no extractability) reduces proof size by ~65%, prover key size by ~75%
- Smallest compressing proofs start around 165 KiB (binding) and 668 KiB (extractable) — recursion threshold



Lattice-based Succinct Arguments from Vanishing Polynomials

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CRYPTO, Santa Barbara, CA, U.S., 2023

Lattice-based Succinct Arguments

Approach	Publicly verifiable	Sublinear-verifier (preprocessing)	Linear-prover
PCP/IOP + linear-only enc. [BCIOP13; BISW17; BISW18; GMNO18]	✗	✓	✓
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Our Results

† New assumption: Vanishing Short Integer Solution (vSIS)

‡ generalization of SIS

† New tool: vSIS commitment for committing to polynomials with short coefficients

‡ Very small ($\text{polylog}(|\text{stmt}|)$) commitment key

‡ (Almost) additively and *multiplicatively* homomorphic

‡ Admit $\tilde{O}(|\text{stmt}|)$ -prover $\text{polylog}(|\text{stmt}|)$ -verifier arguments for commitment openings

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Instantiations	$ \pi $	$\text{Time}(\mathcal{P})$	$\text{Time}(\mathcal{V})$	Setup	Assumptions
Folding	$\tilde{O}_\lambda(1)$	$\tilde{O}_\lambda(\text{stmt})$	$\tilde{O}_\lambda(1)$	Transparent	vSIS (+ RO for NI)
Knowledge assumption	$\tilde{O}_\lambda(1)$	$\tilde{O}_\lambda(\text{stmt})$	$\tilde{O}_\lambda(1)$	Trusted	vSIS + Knowledge-kRISIS

Roadmap

1. vSIS assumptions and commitments
2. Quadratic Relations using vSIS commitments
3. Succinct arguments for vSIS commitment openings

Short Integer Solution (SIS) Assumption

† Parameters: # rows n , # columns m , modulus q .

† Instance: A matrix $\mathbf{A} \in \mathcal{R}_q^{n \times m}$.

† Problem: Find a short vector $\mathbf{u} \in \mathcal{R}^m$ such that

$$\mathbf{A} \cdot \mathbf{u} = \mathbf{0} \pmod{q} \quad \text{and} \quad 0 < \|\mathbf{u}\| \approx 0.$$

† Shorthand: If \mathbf{u} is a short non-zero vector satisfying $\mathbf{A} \cdot \mathbf{u} = \mathbf{v} \pmod{q}$, write

$$\mathbf{u} \in \mathbf{A}^{-1}(\mathbf{v}).$$

Vanishing SIS as SIS Generalisations

SIS

Find short solution to linear equations

$$\mathbf{A} \cdot \mathbf{u} = \mathbf{0} \text{ mod } q \quad \text{and} \quad 0 < \|\mathbf{u}\| \approx 0.$$

SIS (Alternative Interpretation)

Find linear function with short coefficients which vanishes at all given points

Vanishing SIS (vSIS)

Find polynomial (from some class) with short coefficients which vanishes at all given points

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Vanishing Short Integer Solution (vSIS) Assumption

Example: Univariate

† Problem: Find short degree m polynomial without constant term

$$p(X) = p_1X + \dots + p_mX^m \in \mathcal{R}[X]$$

which vanishes at $v \in \mathcal{R}_q^\times$ modulo q , i.e.

$$p(v) = 0 \pmod{q} \quad \text{and} \quad 0 < \|p\| \approx 0.$$

In other words, find short vector $\mathbf{p} \in \mathcal{R}^m$ such that

$$[v \quad v^2 \quad \dots \quad v^m] \cdot \mathbf{p} = 0 \pmod{q} \quad \text{and} \quad 0 < \|\mathbf{p}\| \approx 0.$$

Simple vSIS Commitments (or Hash Functions)

† Domain: Polynomials $p \in \mathcal{R}[X, X^{-1}]$ (of some class) with short coefficients.

† Public parameters: Random unit $v \leftarrow_{\$} \mathcal{R}_q^\times$.

† Commitment of polynomial p :

$$\text{com}(p) = p(v) \bmod q.$$

† Binding: If $p(v) = p'(v) \bmod q$, then we break vSIS, i.e.

$$(p - p')(v) = 0 \bmod q \qquad \|p - p'\| \leq \|p\| + \|p'\| \approx 0.$$

† (Almost) additively and multiplicatively homomorphic (w.r.t. polynomial addition and multiplications):

$$\begin{aligned} p(v) + p'(v) &= (p + p')(v) \bmod q & \|p + p'\| &\leq \|p\| + \|p'\| \approx 0 \\ p(v) \cdot p'(v) &= (p \cdot p')(v) \bmod q & \|p \cdot p'\| &\lesssim \|p\| \cdot \|p'\| \approx 0. \end{aligned}$$

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Encoding Vectors as (Laurent) Polynomials

$$\begin{aligned}
 \mathbf{a} &:= (a_1, \dots, a_m) \in \mathcal{R}^m & \bar{\rho}_{\mathbf{a}}(X) &:= \rho_{\mathbf{a}}(X^{-1}) := a_1 X^{-1} + a_2 X^{-2} + \dots + a_m X^{-m} \\
 \mathbf{b} &:= (b_1, \dots, b_m) \in \mathcal{R}^m & \rho_{\mathbf{b}}(X) &:= b_1 X + b_2 X^2 + \dots + b_m X^m
 \end{aligned}$$

Note that

$$\bar{\rho}_{\mathbf{a}}(X) \cdot \rho_{\mathbf{b}}(X) = \hat{\rho}_{\mathbf{a} * \mathbf{b}}(X) \implies \hat{\rho}_{\mathbf{a} * \mathbf{b}} \text{ has } O(m) \text{ terms (lots of collisions!)}$$

where

$$\dagger \mathbf{a} * \mathbf{b} := \left(\sum_{j-i=k} a_i \cdot b_j \right)_{k=-m}^m \text{ "convolution", and}$$

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Key Example

Want to prove that \mathbf{x} is binary (i.e. $x_i \cdot (1 - x_i) = 0$ for all i).

† \mathbf{x} is committed in vSIS commitment as $c_{\mathbf{x}} := p_{\mathbf{x}}(v)$.

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To prove that a vSIS commitment is committing to a (Laurent) polynomial without constant term:

$$\begin{bmatrix} v & v^2 & \dots & v^m \\ v^{-1} & v^{-2} & \dots & v^{-m} \end{bmatrix} \cdot \mathbf{x} = \begin{bmatrix} c_{\mathbf{x}} \\ \bar{c}_{\mathbf{x}} \end{bmatrix} \bmod q \wedge \|\mathbf{x}\| \approx 0,$$

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Knowledge-kRISIS Assumption(s) [ACLMT22] (a Member of)

† Parameters:

- ‡ SIS parameters (n, m, q) ,
- ‡ submodule rank $t < n$, and
- ‡ t -tuples of Laurent monomials \mathcal{G} .

† Assumption: If a PPT (quantum) algorithm \mathcal{A} , which on input

$$(\mathbf{A}, \mathbf{T}, v, (\mathbf{u}_g)_{g \in \mathcal{G}})$$

where $\mathbf{A} \in \mathcal{R}_q^{n \times m}$, $\mathbf{T} \in (\mathcal{R}_q^\times)^{n \times t}$, $v \in \mathcal{R}_q^\times$, and $\mathbf{u}_g \in \mathbf{A}^{-1}(\mathbf{T} \cdot \mathbf{g}(v))$,

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Want to prove \hat{c} and $\mathbf{w} \in \mathcal{R}^{2m+1}$ satisfies:

$$w_0 = 0$$

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Lattice-based Bulletproofs

Goal: Prove SIS relation with $O(\log m)$ communication:

$$\mathbf{x} \in \mathcal{R}^m : \mathbf{M} \cdot \mathbf{x} = \mathbf{y} \bmod q \wedge \|\mathbf{x}\| \approx 0$$

where $m = 2^\ell$, $\mathbf{M} = [\mathbf{M}_1 \mid \mathbf{M}_2]$, $\mathbf{x} = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix}$.

Prover $\mathcal{P}((\mathbf{M}, \mathbf{y}), \mathbf{x})$

$$\mathbf{y}_{12} := \mathbf{M}_1 \cdot \mathbf{x}_2$$

$$\mathbf{y}_{21} := \mathbf{M}_2 \cdot \mathbf{x}_1$$

$$\hat{\mathbf{x}}_c := c \cdot \mathbf{x}_1 + \mathbf{x}_2$$

$$\xrightarrow{\mathbf{y}_{12}, \mathbf{y}_{21}}$$

$$\xleftarrow{c}$$

$$\xrightarrow{\hat{\mathbf{x}}_c}$$

Verifier $\mathcal{V}(\mathbf{M}, \mathbf{y})$

$$c \leftarrow \$ \mathcal{C}$$

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Just another SIS relation but with only $m/2$ columns \implies Recursion

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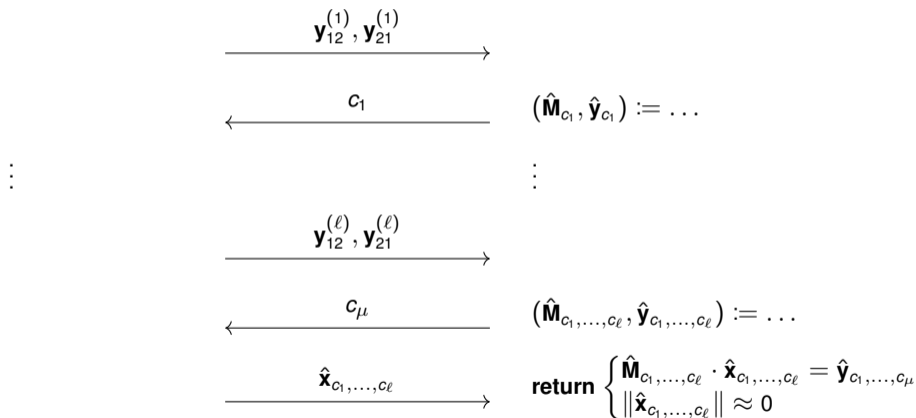
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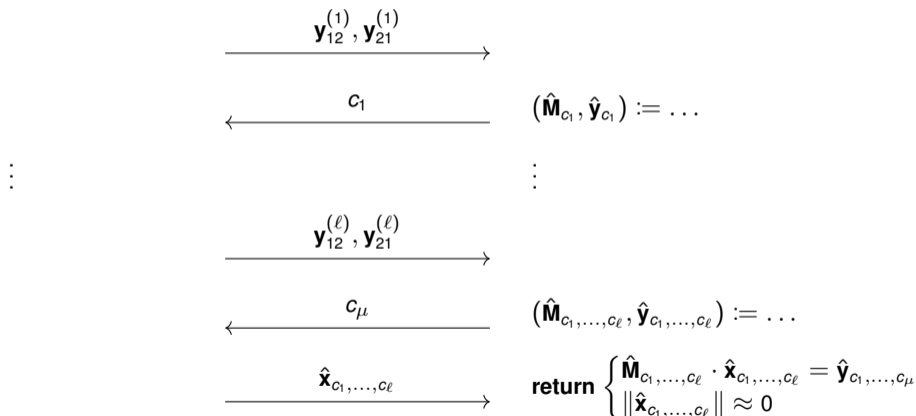
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Structured Folding for vSIS

Core Idea

For \mathbf{M} corresponding to vSIS instance, computing $\hat{\mathbf{M}}_{c_1, \dots, c_\ell}$ takes $\tilde{O}_\lambda(\log m) = \tilde{O}_\lambda(1)$ time.

Example for $\ell = 3$

$$\begin{aligned} \mathbf{M} &= (v \quad v^2 \quad v^3 \quad v^4 \quad v^5 \quad v^6 \quad v^7 \quad v^8) \\ \hat{\mathbf{M}}_{c_1} &= (v \quad v^2 \quad v^3 \quad v^4) + (v^5 \quad v^6 \quad v^7 \quad v^8) \cdot c_1 \\ &= (v \quad v^2 \quad v^3 \quad v^4) \cdot (1 + v^4 \cdot c_1) \\ \hat{\mathbf{M}}_{c_1, c_2} &= (v \quad v^2) \cdot (1 + v^4 \cdot c_1) \cdot (1 + v^2 \cdot c_2) \\ \hat{\mathbf{M}}_{c_1, c_2, c_3} &= v \cdot (1 + v^4 \cdot c_1) \cdot (1 + v^2 \cdot c_2) \cdot (1 + v \cdot c_3) \\ &= v \cdot \prod_{i=1}^3 (1 + v^{2^{3-i}} \cdot c_i) \end{aligned}$$

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Conclusion

- † Vanishing Short Integer Solution (vSIS) assumption and commitments
- † Succinct arguments for vSIS commitment openings
- † Used to construct succinct arguments for NP
 - ‡ Lattice-based
 - ‡ Quasi-linear-time prover
 - ‡ Public and Polylogarithmic-time verifier (after preprocessing)
 - ‡ Transparent setup (RO instantiation)

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