

Worst-Case Lattice Sampler with Truncated Gadgets and Applications

March 19th, 2025

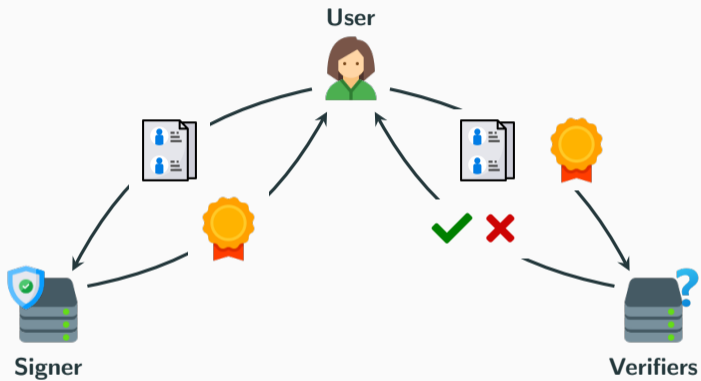
Corentin Jeudy

Orange, Applied Crypto Group

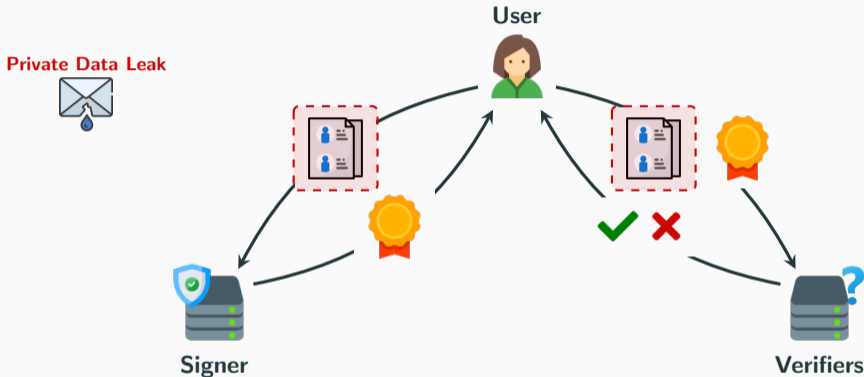


Joint work with Olivier Sanders

Digital Signatures

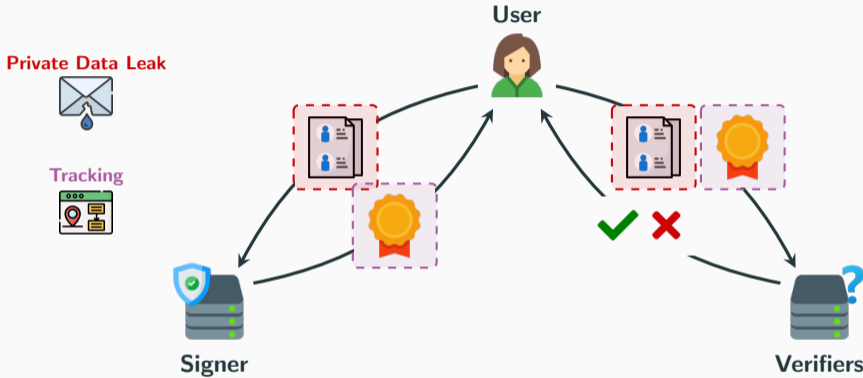


Digital Signatures



No control over the disclosed information: Verifiers (and attacker) learn everything

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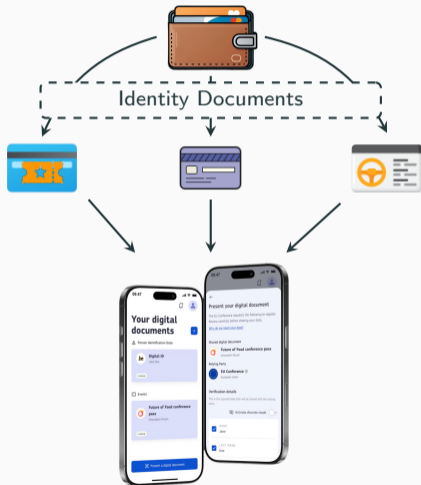


No control over the disclosed information: Verifiers (and attacker) learn everything
Traceable across different authentications: Same signature allows tracing

Privacy as Positive Differentiator in Use-Cases: Digital Identity

European Digital Identity (EUDI) Wallet initiative

"a safe, reliable, and private means of digital identification for everyone in Europe."



Emphasis on

- ✓ Anonymity
- ✓ Unlinkability
- ✓ Selective disclosure



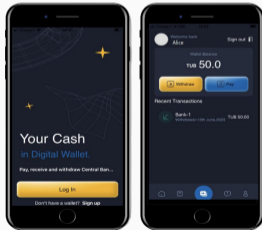
Anonymous
Credentials

Digital Euro initiative (ECB) & Project Tourbillon (BIS & SNB)
"would not identify you or track your payments [...] for cash-like privacy"



Emphasis on

- ✓ (Payer) Anonymity
- ✓ Unlinkability
- ✓ Scalability

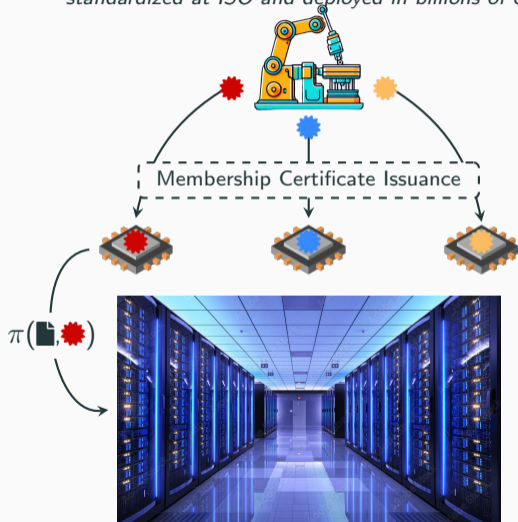


Blind
Signature



eCash

Group Attestation with Built-in Revocation Mechanisms *"standardized at ISO and deployed in billions of chips (TPM, Intel)"*



Emphasis on

- ✓ Anonymity
- ✓ Unlinkability
- ✓ Revocability



Anonymous
Credentials



eCash



EPID

(group signatures)



How is **privacy** usually obtained? **Zero-Knowledge Proof of Signature & Message**





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Proof of x
s.t. $g^x = h$

Proof of x
s.t. $Ax = u \wedge \|x\| \leq B$

Proof of x
s.t. $\mathcal{H}(x) = h$

Algebraic

Generic

Privacy from Zero-Knowledge Proofs



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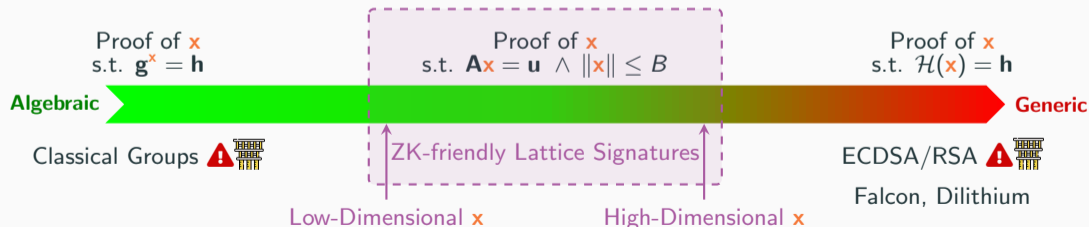
ECDSA/RSA  

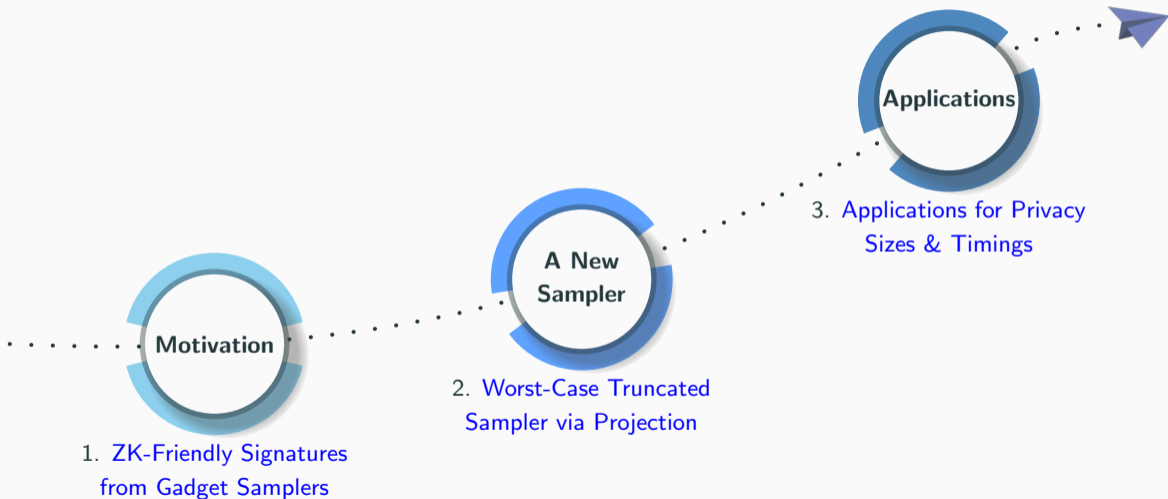
Falcon, Dilithium

Privacy from Zero-Knowledge Proofs



How is **privacy** usually obtained? **Zero-Knowledge Proof of Signature & Message**







Zero-Knowledge-Friendly Signatures from Gadget Samplers

ISIS _{m, d, q, β}

Given $(\mathbf{A}, \mathbf{u}) \leftarrow U(R_q^{d \times m+1})$, find $\mathbf{x} \in R^m$ such that $\mathbf{Ax} = \mathbf{u} \pmod q$, $\|\mathbf{x}\| \leq \beta$.

When $\mathbf{u} = \mathbf{0}$, we ask $\mathbf{x} \neq \mathbf{0}$.

Decision: Distinguish $\mathbf{Ax} \pmod q$ for a random short \mathbf{x} from a random $\mathbf{u} \rightarrow$ **LWE**.

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ISIS is hard unless we know a trapdoor \mathbf{R} on \mathbf{A} .

- Ability to invert $f_{\mathbf{A}} : \mathbf{x} \mapsto \mathbf{Ax} \pmod q$ over bounded domain
- Ability to randomize preimage finding without leaking $\mathbf{R} \rightarrow$ **Preimage Sampling**
- Design secure signatures [GPV08]¹: Find short \mathbf{x} such that $\mathbf{Ax} = \mathcal{H}(\mathbf{m}) \pmod q$

¹Gentry, Peikert, Vaikuntanathan. Trapdoors for Hard Lattices and New Cryptographic Constructions. STOC 2008.

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Gadget-based samplers [MP12]¹ are well suited for signatures without ROM



¹Micciancio, Peikert. Trapdoors for Lattices: Simpler, Tighter, Faster, Smaller. Eurocrypt 2012

Micciancio-Peikert trapdoors [MP12]: Family of matrices \mathbf{A}_T such that

$$\mathbf{A}_T = [\mathbf{A}' | \mathbf{T}\mathbf{G} - \mathbf{A}'\mathbf{R}] \text{ and } \mathbf{A}' = [\mathbf{I} | \mathbf{A}]$$


verifies $\mathbf{A}_T \mathbf{L} = \mathbf{T}\mathbf{G} \bmod q$, with $\mathbf{L} = \begin{bmatrix} \mathbf{R} \\ \mathbf{I} \end{bmatrix}$

with $\mathbf{G} = [b^0 \mathbf{I} | \dots | b^{k-1} \mathbf{I}]$, and $k = \log_b q$
(base- b decomposition)

 \mathbf{R}  $\mathbf{B} = \mathbf{A}'\mathbf{R}$
 $\mathbf{T} (= t\mathbf{I})$

Naive Approach: Compute \mathbf{z} so that $\mathbf{T}\mathbf{G}\mathbf{z} = \mathbf{u} \bmod q$, and return $\mathbf{L}\mathbf{z}$ as preimage of \mathbf{u}

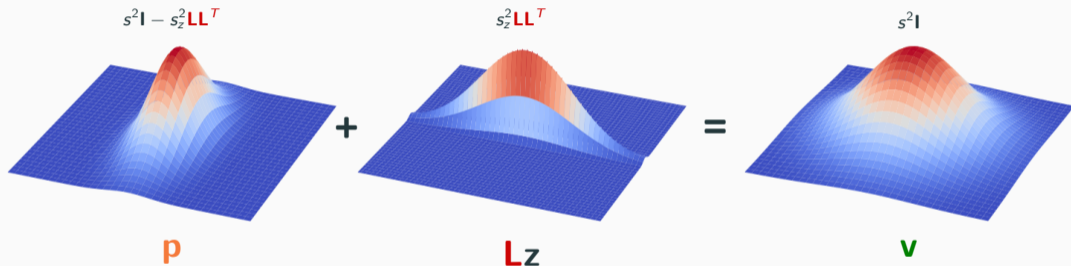
 Collecting many preimages will leak \mathbf{R} ...

 Distribution on \mathbf{z} and add mask \mathbf{p} : preimages $\mathbf{v} = \mathbf{p} + \mathbf{L}\mathbf{z} = \begin{bmatrix} \mathbf{p}_1 + \mathbf{R}\mathbf{z} \\ \mathbf{p}_2 + \mathbf{z} \end{bmatrix}$

(and syndrome correction so that $\mathbf{T}\mathbf{G}\mathbf{z} = \mathbf{w} = \mathbf{u} - \mathbf{A}_T \mathbf{p}$)

How to Choose the Mask? Spherical Convolution

📖 Compensate statistical leakage by adapting covariance of \mathbf{p} [MP12]. Only for \mathbf{z} and \mathbf{p} Gaussian



Quality: $s \approx s_z \sqrt{1 + \|\mathbf{R}\|_2^2}$ with $s_z \approx \eta_\varepsilon(\mathcal{L}_q^\perp(\mathbf{G}))$.

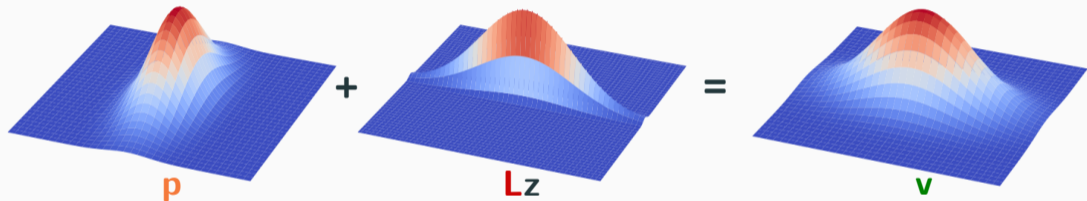
How to Choose the Mask? Elliptical Convolution

💡 Use **elliptical Gaussians** instead of spherical

$$\begin{bmatrix} s_1^2 \mathbf{I} & \\ & s_2^2 \mathbf{I} \end{bmatrix} - s_z^2 \mathbf{L}\mathbf{L}^T$$

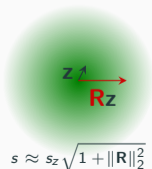
$$s_z^2 \mathbf{L}\mathbf{L}^T$$

$$\begin{bmatrix} s_1^2 \mathbf{I} & \\ & s_2^2 \mathbf{I} \end{bmatrix}$$

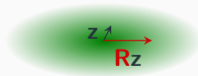


Spherical Sampling

Elliptical Sampling



$$\mathbf{v} = \mathbf{p} + \begin{bmatrix} \mathbf{R}z \\ z \end{bmatrix}$$



$$s_1 \approx \sqrt{2} s_z \|\mathbf{R}\|_2, \quad s_2 \approx \sqrt{2} s_z$$

We let $s_z \approx \eta_\varepsilon(\mathcal{L}_q^\perp(\mathbf{G}))$, $s_1 \approx \sqrt{2}s_z\|\mathbf{R}\|_2$, $s_2 \approx \sqrt{2}s_z$ and define

$$\mathbf{S}_p = \begin{bmatrix} s_1^2 \mathbf{I}_{2d} & \\ & s_2^2 \mathbf{I}_{dk} \end{bmatrix} - s_z^2 \begin{bmatrix} \mathbf{R} \\ \mathbf{I}_{dk} \end{bmatrix} \begin{bmatrix} \mathbf{R}^T & \mathbf{I}_{dk} \end{bmatrix}$$

The sampler finds a preimage of \mathbf{u} for $\mathbf{A}_T = [\mathbf{A}' | \mathbf{T}\mathbf{G} - \mathbf{A}'\mathbf{R}]$

- $\mathbf{p} \leftarrow \mathcal{D}_{\mathbb{Z}^{d(2+k)}, \sqrt{s_p}}$
- $\mathbf{w} \leftarrow \mathbf{T}^{-1}(\mathbf{u} - \mathbf{A}_T \mathbf{p}) \bmod q$
- $\mathbf{z} \leftarrow \mathcal{D}_{\mathcal{L}_q^\mathbf{w}(\mathbf{G}), s_z}$
- $\begin{bmatrix} \mathbf{v}_1 \\ \mathbf{v}_2 \end{bmatrix} \leftarrow \mathbf{p} + \begin{bmatrix} \mathbf{R} \\ \mathbf{I} \end{bmatrix} \mathbf{z}$
- Output $\mathbf{v} = (\mathbf{v}_1, \mathbf{v}_2)$

verifies $\mathbf{G}\mathbf{z} = \mathbf{w} \bmod q$

$$= \mathbf{p} + \mathbf{L}\mathbf{z}$$

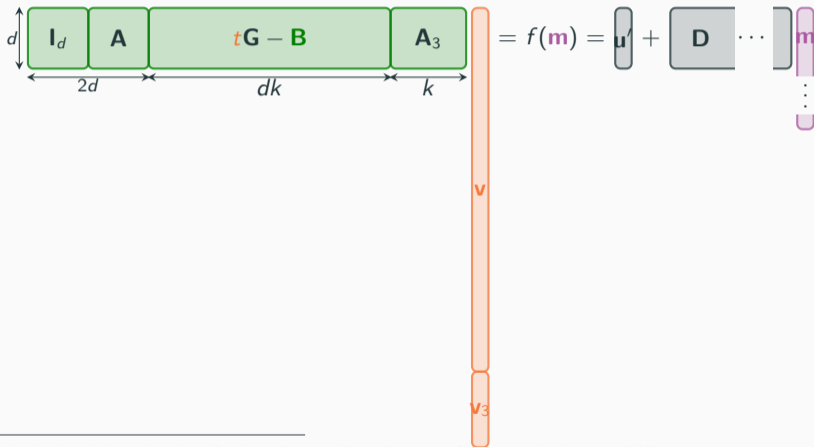
verifies $\mathbf{A}_T \mathbf{v} = \mathbf{u} \bmod q$

MP Sampler

ZK-Friendly Signature from Gadget Sampler

Signature scheme from [AGJ+24]²:

🔑 : \mathbf{R}
 🔑 : $\mathbf{B} = [\mathbf{I}_d | \mathbf{A}] \mathbf{R}$
 💡 : $t, \mathbf{v}, \mathbf{v}_3$
 📄 : \mathbf{m}
 PP : $(\mathbf{A}, \mathbf{A}_3, \mathbf{D}, \mathbf{u}', \mathbf{G} = [b^0 \mathbf{I} | \dots | b^{k-1} \mathbf{I}])$

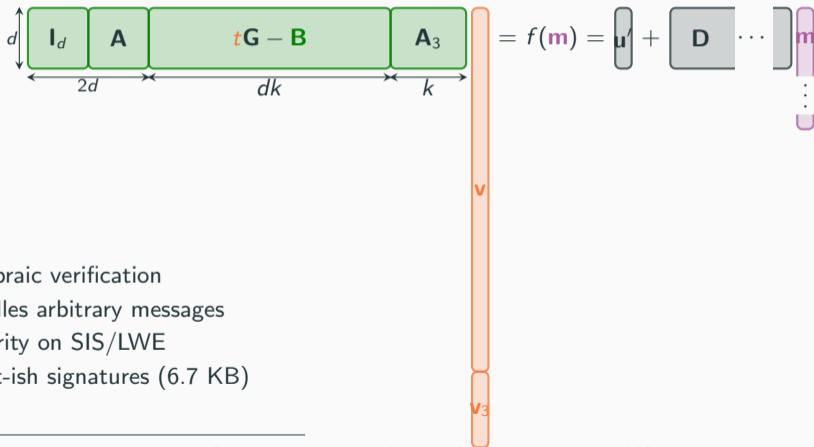


²Argo, Güneysu, Jeudy, Land, Roux-Langlois, Sanders. Practical Post-Quantum Signatures for Privacy. CCS 2024

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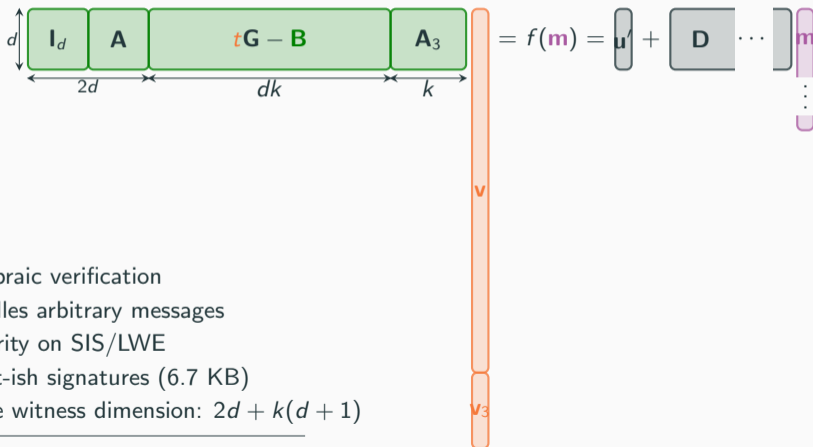
- ✓ Algebraic verification
- ✓ Handles arbitrary messages
- ✓ Security on SIS/LWE
- ✓ Short-ish signatures (6.7 KB)

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- ✓ Algebraic verification
- ✓ Handles arbitrary messages
- ✓ Security on SIS/LWE
- ✓ Short-ish signatures (6.7 KB)
- ✗ Large witness dimension: $2d + k(d + 1)$

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Worst-Case Sampler with Truncated Gadgets via Projection

 Reduce gadget dimension with “approximate trapdoors” [CGM19]³: Sampling \mathbf{v}' s.t. $\mathbf{A}'_{\mathbf{T}}\mathbf{v}' + \mathbf{e} = \mathbf{u}$ with \mathbf{e} small is sufficient.

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Note $\mathbf{G}_L = [b^0\mathbf{I}_d | \dots | b^{\ell-1}\mathbf{I}_d]$, $\mathbf{G}_H = [b^\ell\mathbf{I}_d | \dots | b^{k-1}\mathbf{I}_d]$. Now: $\mathbf{A}'_{\mathbf{T}} = [\mathbf{A}' | \mathbf{T}\mathbf{G}_H - \mathbf{A}'\mathbf{R}]$, with $\mathbf{A}' = [\mathbf{I}_d | \mathbf{A}]$.

$$\mathbf{A}'_{\mathbf{T}}\mathbf{v}' + \mathbf{e} = \mathbf{u} \iff \underbrace{[\mathbf{I}_d | \mathbf{A} | \mathbf{T}\mathbf{G}_H - \mathbf{A}'\mathbf{R}]}_{\text{exact preimage}} \begin{pmatrix} \mathbf{v}' + \begin{bmatrix} \mathbf{e} \\ \mathbf{0} \\ \mathbf{0} \end{bmatrix} \end{pmatrix} = \mathbf{u}$$

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Can we handle the **convolution as before** with the **additional error \mathbf{e}** ?

³Chen, Genise, Mukherjee. Approximate trapdoors for lattices and smaller hash-and-sign signatures. Asiacrypt 2019.

What About Security?



Reduced dimension, but what about security?



Reduced dimension, but what about security? **Well, it's complicated.**

⊗ To prove \mathbf{v} does not leak \mathbf{R} , [CGM19] must be able to simulate \mathbf{e} (as it depends on \mathbf{p}). Requires knowing the distribution of \mathbf{e} , which causes two problems:

- ① Distribution of \mathbf{e} difficult when \mathbf{u} is arbitrary/adversarially chosen
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≈ Proposed solution requires $\mathbf{u} = f(\mathbf{m})$ to be a *consistent, random, reprogrammable* function of \mathbf{m} . That is... a **random oracle**.

✓ Fine for hash-and-sign standard signatures,

✗ Not for ZK-friendly signatures, where $f(\mathbf{m})$ is algebraic (e.g. $f(\mathbf{m}) = \mathbf{u}' + \mathbf{Dm}$).

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[CGM19] **not applicable to the main use-cases** of gadget samplers (\mathbf{u} arbitrary)

💡 Use the perturbation to hide (some of) the error using convolution. Split \mathbf{R} into $(\mathbf{R}_1, \mathbf{R}_2)$ so that $[\mathbf{I}_d | \mathbf{A}] \mathbf{R} = \mathbf{R}_1 + \mathbf{A} \mathbf{R}_2$. The unperturbed preimage is

$$\mathbf{v} = \begin{bmatrix} \mathbf{R}_1 \\ \mathbf{R}_2 \\ \mathbf{I}_{d(k-\ell)} \end{bmatrix} \mathbf{z}_H + \begin{bmatrix} \mathbf{T} \mathbf{G}_L \mathbf{z}_L \\ \mathbf{0} \\ \mathbf{0} \end{bmatrix}$$

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- ⊗ \mathbf{G}_L large compared to $\mathbf{R}_i \implies$ needs large perturbation
- ⊗ Matrix not full rank when $\ell > 1 \implies$ complex lattice smoothing analysis

Back To Square One

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factor

$$\begin{bmatrix} \mathbf{G}_L & & \\ & \mathbf{I}_d & \\ & & \mathbf{I}_{d(k-\ell)} \end{bmatrix} \cdot \begin{bmatrix} \mathbf{T} & \mathbf{0} & \mathbf{R}_1 \\ \mathbf{0} & \mathbf{I}_{\ell-1} \otimes \mathbf{T} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{R}_2 \\ \mathbf{0} & \mathbf{0} & \mathbf{I}_{d(k-\ell)} \end{bmatrix}$$

Public Part
(\mathbf{K} -projection)Private Part
(\mathbf{L} full rank)

\mathbf{K} \mathbf{L}

💡 Perturb \mathbf{Lz} and project with \mathbf{K} afterwards.

We need to compensate the covariance $s_z^2 \mathbf{L}\mathbf{L}^T$

$$\mathbf{L}\mathbf{L}^T = \begin{bmatrix} \mathbf{T}\mathbf{T}^T + \mathbf{R}_1\mathbf{R}_1^T & \mathbf{0} & \mathbf{R}_1\mathbf{R}_2^T & \mathbf{R}_1 \\ \mathbf{0} & \mathbf{I}_{\ell-1} \otimes \mathbf{T}\mathbf{T}^T & \mathbf{0} & \mathbf{0} \\ \mathbf{R}_2\mathbf{R}_1^T & \mathbf{0} & \mathbf{R}_2\mathbf{R}_2^T & \mathbf{R}_2 \\ \mathbf{R}_1^T & \mathbf{0} & \mathbf{R}_2^T & \mathbf{I}_{d(k-\ell)} \end{bmatrix}$$

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💡 We aim for $\mathbf{S} = \text{diag}(s_1^2, s_2^2, s_3^2, s_4^2)$. We expect to need

$$s_1 = O(s_z(\|\mathbf{T}\|_2 + \|\mathbf{R}_1\|_2)), \quad s_2 = O(s_z\|\mathbf{T}\|_2), \quad s_3 = O(s_z\|\mathbf{R}_2\|_2) \quad \text{and} \quad s_4 = O(s_z).$$

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✅ We get $s_1 = \alpha\sqrt{\|\mathbf{T}\|_2^2 + 3\|\mathbf{R}_1\|_2^2}$, $s_2 = \alpha\|\mathbf{T}\|_2$, $s_3 = \alpha\sqrt{3}\|\mathbf{R}_2\|_2$ and $s_4 = \alpha\sqrt{3}$ are sufficient, with $\alpha = s_z^2 / \sqrt{s_z^2 - \eta_\varepsilon(\mathbb{Z}^{dk})^2} \approx s_z$.

We then take $\mathbf{A}_T = [\mathbf{A}' | \mathbf{T}\mathbf{G}_H - \mathbf{A}'\mathbf{R}]$ and

$$\mathbf{S} = \begin{bmatrix} s_1^2 \mathbf{I}_d & & & \\ & s_2^2 \mathbf{I}_{d(\ell-1)} & & \\ & & s_3^2 \mathbf{I}_d & \\ & & & s_4^2 \mathbf{I}_{d(k-\ell)} \end{bmatrix}$$

- $\mathbf{p} \leftarrow \mathcal{D}_{\mathbb{Z}^{d(k+1)}, \sqrt{\mathbf{S}_p}}$

$$\mathbf{S}_p = \mathbf{S} - s_z^2 \mathbf{L}\mathbf{L}^T$$

- $\mathbf{w} \leftarrow \mathbf{T}^{-1}(\mathbf{u} - \mathbf{A}_T \mathbf{K} \mathbf{p}) \bmod q$

verifies $\mathbf{G}\mathbf{z} = \mathbf{w} \bmod q$

- $\mathbf{z} \leftarrow \mathcal{D}_{\mathcal{L}_q^{\mathbf{w}}(\mathbf{G}), s_z}$

- $\mathbf{v}' \leftarrow \mathbf{p} + \mathbf{L}\mathbf{z}$

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- Output $\mathbf{v} = \mathbf{K}\mathbf{v}'$

Truncated Sampler

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Truncated Sampler



Let us zoom in on the **perturbation sampler**

Perturbation sampling represents the vast majority of the computation time. Let's optimize with precomputations. Take $\mathbf{T} = t\mathbf{I}_d$ with t invertible modulo q .

$$\mathbf{S}_p = \begin{bmatrix} s_1^2 \mathbf{I} - s_z^2 (tt^* \mathbf{I}_d + \mathbf{R}_1 \mathbf{R}_1^*) & \mathbf{0} & -s_z^2 \mathbf{R}_1 \mathbf{R}_2^* & -s_z^2 \mathbf{R}_1 \\ \mathbf{0} & s_2^2 \mathbf{I} - s_z^2 tt^* \mathbf{I}_{d(\ell-1)} & \mathbf{0} & \mathbf{0} \\ -s_z^2 \mathbf{R}_2 \mathbf{R}_1^* & \mathbf{0} & s_3^2 \mathbf{I} - s_z^2 \mathbf{R}_2 \mathbf{R}_2^* & -s_z^2 \mathbf{R}_2 \\ -s_z^2 \mathbf{R}_1^* & \mathbf{0} & -s_z^2 \mathbf{R}_2^* & s_4^2 \mathbf{I} - s_z^2 \mathbf{I}_{d(k-\ell)} \end{bmatrix}$$

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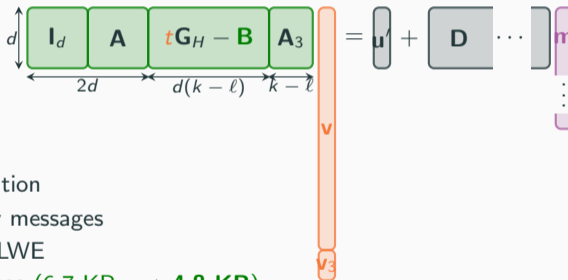
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- ② Part in s_3^2 and s_4^2 independent of t . Precomputation done at key generation
- ③ Part in s_1^2 depends on t . Schur complements must be computed online. But only d dimensions out of $d(k+1)$



Applications:
(More) Practical Post-Quantum Privacy

Signature in the Standard Model

🔑 : R_1, R_2
 🔑 : $B = R_1 + AR_2$
 💡 : t, v, v_3
 📄 : m
 PP : $(A, A_3, D, u', G_H = [b^\ell I \mid \dots \mid b^{k-1} I])$



- ✓ Algebraic verification
- ✓ Handles arbitrary messages
- ✓ Security on SIS/LWE
- ✓ **Shorter signatures** (6.7 KB \rightarrow 4.8 KB)
- ✓ **Smaller witness dimension:** $2d + k(d + 1) \rightarrow 2d + (k - \ell)(d + 1)$

Signature in the Standard Model: Performance

For $k = 5$:

	$ \text{pk} $	$ \text{sig} $	Sec. (Core-SVP)
$\ell = 0$	47.5 KB	6.7 KB	126
$\ell = 1$	38.0 KB	5.9 KB	123
$\ell = 2$	28.5 KB	4.8 KB	121

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Procedure	Average Time ($\ell = 0$)	Average Time ($\ell = 2$)
SamplePerturb	52.0 ms	80.2 ms
SampleGadget	1.8 ms	1.8 ms
SamplePre	56.5 ms	83.9 ms
Sign	56.9 ms	84.3 ms
Verify	1.1 ms	0.7 ms

Small overhead due to online covariance computations

Example improvements in **group signatures** [LNPS21]⁴ [LNP22]⁵, **anonymous credentials** [AGJ⁺24]⁶, **blind signatures** [JS24]⁷

	Improvement	Final Size
Group Signature	15.7 %	$ \text{gsig} = 75.7 \text{ KB}$
Anonymous Credentials	11.2 %	$ \text{show} = 54.0 \text{ KB}$
Blind Signature	11.8 %	$ \text{bsig} = 36.3 \text{ KB}$

(Full comparison in the paper (2024/1952), with different values of ℓ)

⁴Lyubashevsky, Nguyen, Plançon, Seiler. Shorter Lattice-Based Group Signatures via “Almost Free” Encryption and Other Optimizations. Asiacrypt 2021

⁵Lyubashevsky, Nguyen, Plançon. Lattice-Based Zero-Knowledge Proofs and Applications: Shorter, Simpler and More General. Crypto 2022

⁶Argo, Güneysu, Jeudy, Land, Roux-Langlois, Sanders. Practical Post-Quantum Signatures for Privacy. CCS 2024

⁷Jeudy, Sanders. Improved Lattice Blind Signatures from Recycled Entropy. ePrint 2024/1289

Conclusion and Directions



Preimage Sampler with Truncated Gadgets in the **worst case**

- › Unlocks truncated gadgets in their main applications
- › Same structure: drop-in replacement to full gadget sampler [[MP12](#)]
- › Reduced dimension: immediate improvement in many privacy-driven applications



Perspectives



More efficient perturbation sampler?






Optimized implementation (dedicated backend, parallelization, parameter selection)

- ✓ **Preimage Sampler with Truncated Gadgets** in the **worst case**
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- ? **Perspectives**
 - 🧱 More efficient perturbation sampler?
 - 🔧 Optimized implementation (dedicated backend, parallelization, parameter selection)

Thank You!

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