

PQC + Threshold

State of the Art in Threshold Quantum-Resistant Signatures

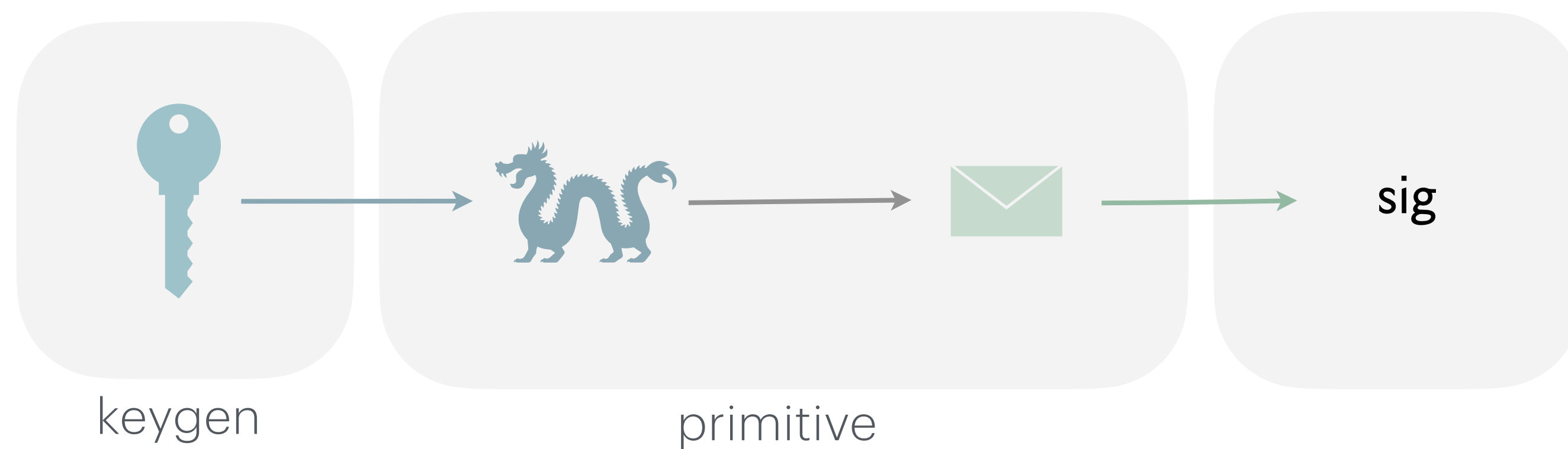
Guilhem Niot

CryptoDay @ Télécom Paris – 18/09/2025



Threshold Signatures

Centralized setting



Threshold Signatures

What if the party is corrupted or becomes unresponsive...

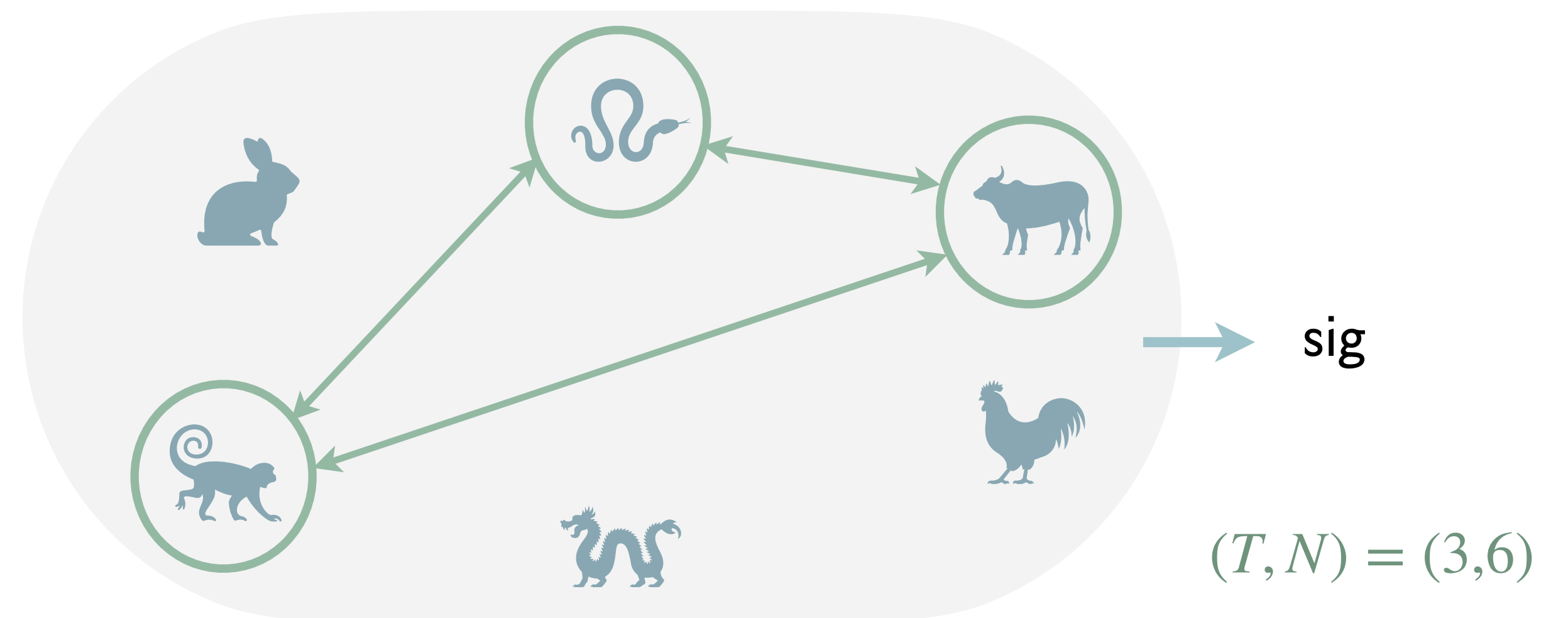
Question: can we split the trust among several parties?

Threshold Signatures

What if the party is corrupted or becomes unresponsive...

Question: can we split the trust among several parties?

Interactive protocol to distribute the scheme:
 T -out-of- N parties can collaborate to sign and
 $T - 1$ parties cannot.



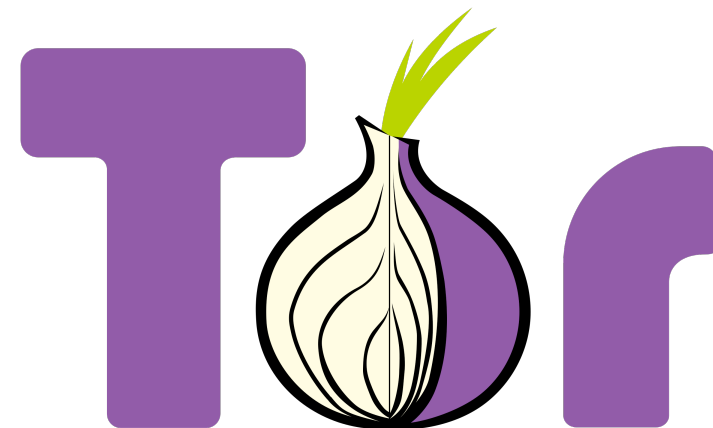
Applications of Threshold Signatures



Cryptocurrency wallets & DeFi



Distributed signing for CDNs



Distributed consensus in Tor

NIST Call for Threshold Schemes

PUBLICATIONS

NIST IR 8214C (2nd Public Draft)

NIST First Call for Multi-Party Threshold Schemes



Date Published: March 27, 2025

Comments Due: April 30, 2025

Email Comments to: nistir-8214C-comments@nist.gov

Author(s)

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Announcement

This is a second public draft. Threshold schemes should NOT be submitted until the final version of this report is published. However, the present draft can be used as a baseline to prepare for future submissions.

The scope of the call is organized into categories related to signing (Sign), public-key encryption (PKE), symmetric-key cryptography and hashing (Symm), key generation (KeyGen), fully homomorphic encryption

Goal of this talk

- Provide an overview of the most practical PQ threshold signatures
- Explain their technical and practical differences

The trade-offs of threshold schemes

Select a base scheme

Lattice-based

Hash-based

Multivariate-based

Isogeny-based

The trade-offs of threshold schemes

Select a base scheme

Lattice-based

Hash-based

Multivariate-based

Isogeny-based

trade-off

Speed

Rounds

Communication

Gap #corruptions / #signers

efficiency

Distributed Key Generation
(DKG)

Detecting Malicious Parties:
Identifiable Aborts (IA)

Backward compatibility

advanced
properties

Post-Quantum Threshold Signatures? Lattice-based

Threshold ML-DSA

Threshold Signatures Reloaded
ML-DSA and Enhanced Raccoon with Identifiable
Aborts

Giacomo Borin¹, Sofia Celi², Rafael del Pino³, Thomas Espitau³, Guilhem Niot^{3,4},
Thomas Prest³

Efficient, Scalable Threshold ML-DSA Signatures: An MPC Approach

Alexander Bienstock^{*‡}, Leo de Castro^{*†✉}, Daniel Escudero^{*‡}, Antigoni Polychroniadou^{*‡}, Akira Takahashi^{*‡}
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- + Standard
- + DKG
- Limited scalability

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FN-DSA based?

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FN-DSA based?

Raccoon based

Threshold Raccoon: Practical Threshold Signatures
from Standard Lattice Assumptions

Rafael del Pino¹, Shuichi Katsumata^{1,2}, Mary Maller^{1,3}, Fabrice Mouhartem⁴, Thomas
Prest¹, Markku-Juhani Saarinen^{1,5}

Simple and Efficient Lattice Threshold
Signatures with Identifiable Abort

Rafael del Pino¹, Thomas Espitau¹, Guilhem Niot^{1,2}, and Thomas
Prest¹

Two-Round Threshold Signature from
Algebraic One-More Learning with Errors

Thomas Espitau¹, Shuichi Katsumata^{1,2}, Kaoru Takemure^{* 1,2}

- + Efficient and scalable
- + DKG + Abort identification
- Non standard (~10KB sig)

Post-Quantum Threshold Signatures?

Hash-based

Threshold SPHINCS+?

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Hash-based

Threshold SPHINCS+?

Aggregating and thresholdizing hash-based signatures using STARKs

Irakliy Khaburzaniya
Polygon/Meta
irakliy81@gmail.com

Kevin Lewi
Meta
klewi@fb.com

Kostantinos Chalkias
Meta
chalkiaskostas@gmail.com

Harjasleen Malvai
UIUC / IC3
hmalvai2@illinois.edu

- Non-standard
- Large $>100\text{KB}$ signatures

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- Non-standard
- Large >100KB signatures

Turning Hash-Based Signatures into Distributed Signatures and Threshold Signatures

Delegate Your Signing Capability, and Distribute it Among Trustees

John Kelsey^{1,2} , Nathalie Lang³  and Stefan Lucks³ 

- Non-standard
- Stateful
- Small number of parties

Post-Quantum Threshold Signatures?

Multivariate

UOV and MAYO

Share the MAYO: thresholdizing MAYO

Sofia Celi¹, Daniel Escudero², and Guilhem Niot³

+ NIST candidates (UOV and MAYO)

Post-Quantum Threshold Signatures?

Isogeny based

CSl-FiSh based

Threshold Schemes from Isogeny Assumptions

Luca De Feo¹[0000–0002–9321–0773] and Michael Meyer^{2,3★}

- + Efficient
- Non-standard
- Less trusted assumption

Post-Quantum Threshold Signatures?

Focus:

- NIST standard or candidate: **Threshold ML-DSA, UOV, MAYO**
- **Raccoon**: scalable + efficient advanced properties

Threshold MAYO and UOV

MAYO and UOV

UOV

- **Small signatures:** as small as 96 bytes.

MAYO

Parameter set	MAYO_one	MAYO_two	MAYO_three	MAYO_five
security level	1	1	3	5
secret key size	24 B	24 B	32 B	40 B
public key size	1420 B	4912 B	2986 B	5554 B
signature size	454 B	186 B	681 B	964 B

MAYO and UOV

Multivariate Quadratic (MQ) cryptography is based on the assumed hardness of finding a solution to **a system of multivariate quadratic equations** (over a finite field).

The current record mod 31 is solving a system of 22 equations in 22 variables.

$$x + 5x^2 + 3xy = 4 \pmod{7}$$

$$x^2 + 5xy + 5y^2 = 1 \pmod{7}$$

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$$\begin{aligned}x + 5x^2 + 3xy &= 4 \pmod{7} \\ x^2 + 5xy + 5y^2 &= 1 \pmod{7}\end{aligned}$$



Define a multivariate map $\mathcal{P}$
 $\mathcal{P}(\mathbf{x}) = \mathbf{t}$

MAYO and UOV

How to design a signature scheme from MQ?

Add some structure to $\mathcal{P}$: define a secret subspace O such that $\mathcal{P}(O) = 0$.

Signing process:

- 1 Derive a target $\mathbf{t} = H(\text{msg})$. Goal: find $\mathbf{x}$ such that $\mathcal{P}(\mathbf{x}) = \mathbf{t}$.
- 2 Sample a random vector $\mathbf{v}$.
- 3 Search for $\mathbf{o} \in O$ such that $\mathcal{P}(\mathbf{v} + \mathbf{o}) = \mathbf{t}$

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$$= \mathcal{P}(\mathbf{v}) + \underbrace{\mathcal{P}(\mathbf{o})}_{=0} + \underbrace{\Delta_{\mathbf{v}}(\mathbf{o})}_{\text{linear}}$$

Threshold MAYO and UOV

High-level signing process

Compute target $\mathbf{t} = H(\text{msg})$

Solve $\mathbf{A} \cdot \mathbf{x} = \mathbf{y}$, for

- $\mathbf{A}$ a randomized function of the secret
- $\mathbf{y}$ a function of $\mathbf{t}$

Threshold MAYO and UOV

High-level signing process

UOV . Sign(sk, msg) $\rightarrow$ sig

- Compute $\mathbf{t} = H(\text{msg})$
- Sample uniform matrix $\mathbf{V}$
- Derive system $\mathbf{A} = \text{ComputeA}(\text{sk}, \mathbf{V})$
- Derive $\mathbf{y} = \text{ComputeY}(\mathbf{t}, \mathbf{V})$
- Return $\mathbf{x} = \mathbf{A}^{-1} \cdot \mathbf{y}$

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Addition: share all secret values

$$\mathbf{a} = \mathbf{a}_1 + \dots + \mathbf{a}_T$$

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Multiplication

Possible with pre-shared triples (a, b, ab)

Threshold MAYO and UOV

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- Inversion algorithm can be implemented efficiently by revealing a masked matrix A (multiplied on both sides by random matrices)
 1. Compute and reveal $\mathbf{R} \cdot \mathbf{A} \cdot \mathbf{T}$ and
 2. Compute $(\mathbf{R} \cdot \mathbf{A} \cdot \mathbf{T})^{-1}$
 3. Compute $\mathbf{x} = \mathbf{T} \cdot (\mathbf{R} \cdot \mathbf{A} \cdot \mathbf{T})^{-1} \cdot \mathbf{R} \cdot \mathbf{y}$

Lattice-based Threshold Signatures

ML-DSA signatures

ML-DSA . Keygen() $\rightarrow$ sk, vk

- $\text{vk} = \mathbf{A} \cdot \text{sk} + \mathbf{e}$, for sk, $\mathbf{e}$ short

MLWE assumption: vk appears uniformly distributed for $\mathbf{A}$ wide enough (more inputs than outputs)

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To sign: prove knowledge of sk, $\mathbf{e}$, without revealing sk, $\mathbf{e}$. (*Fiat-Shamir type signature*)

1

Prover

Sample short $\mathbf{r}$
 $\mathbf{w} = \mathbf{A} \cdot \mathbf{r}$

Challenger

$\mathbf{w}$

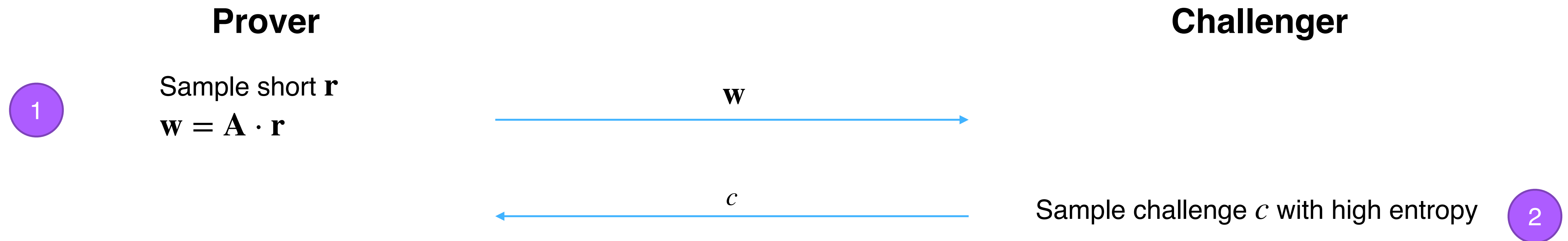
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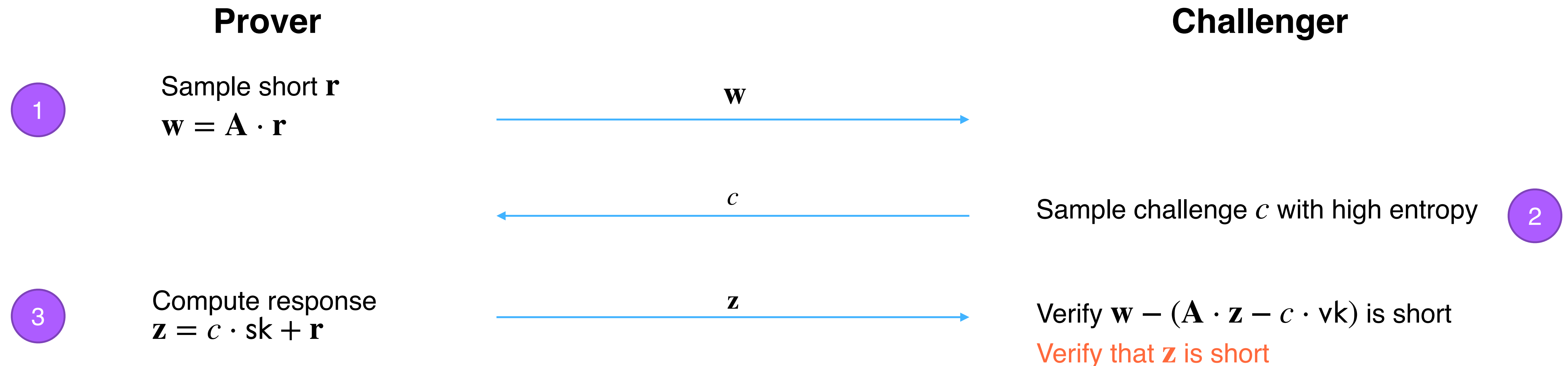
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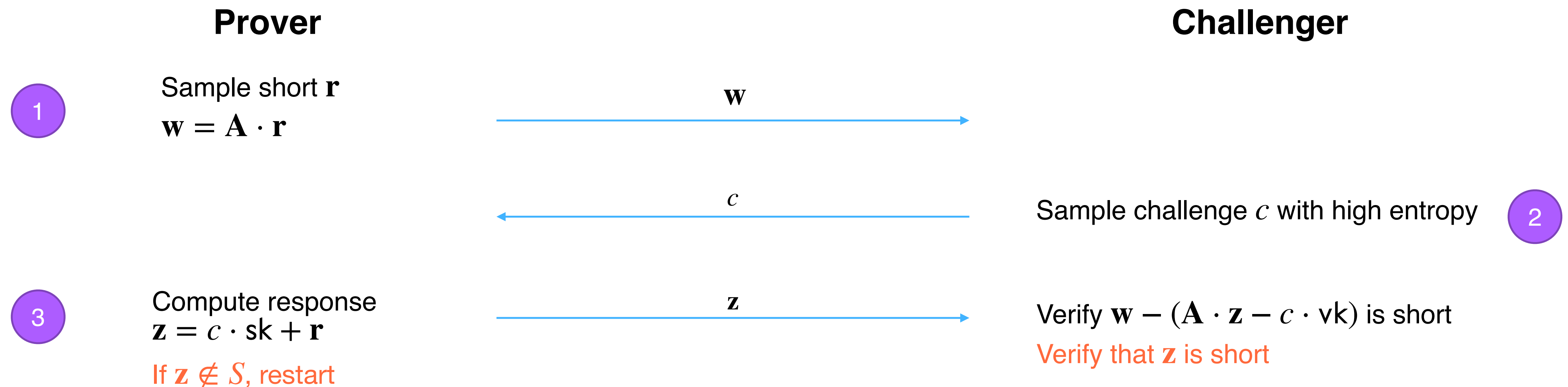
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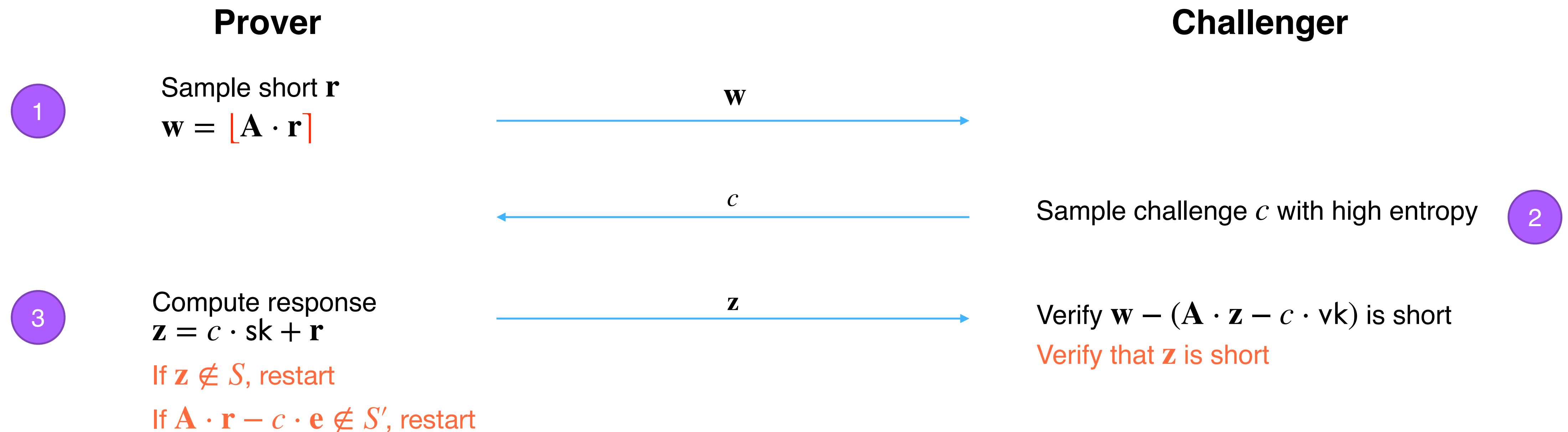
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Prover

1

Sample short $\mathbf{r}$
 $\mathbf{w} = \lfloor \mathbf{A} \cdot \mathbf{r} \rfloor$

2

$c = H(\mathbf{w}, \text{msg})$

3

Compute response
 $\mathbf{z} = c \cdot \text{sk} + \mathbf{r}$

If $\mathbf{z} \notin S$, restart

If $\mathbf{A} \cdot \mathbf{r} - c \cdot \mathbf{e} \notin S'$, restart

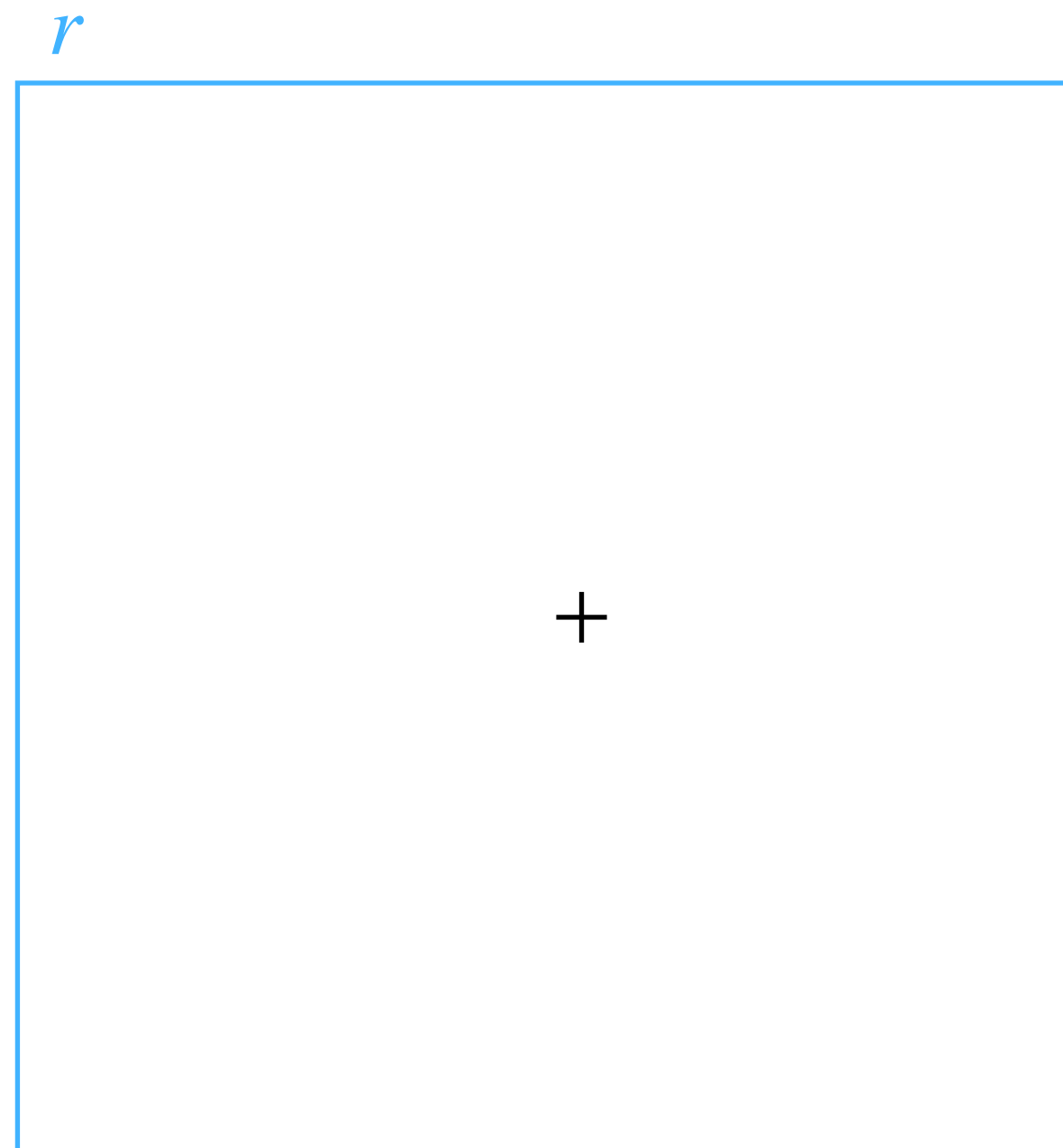
Challenger

$\mathbf{z}$

Verify $\mathbf{w} - (\mathbf{A} \cdot \mathbf{z} - c \cdot \text{vk})$ is short
Verify that $\mathbf{z}$ is short

Rejection sampling

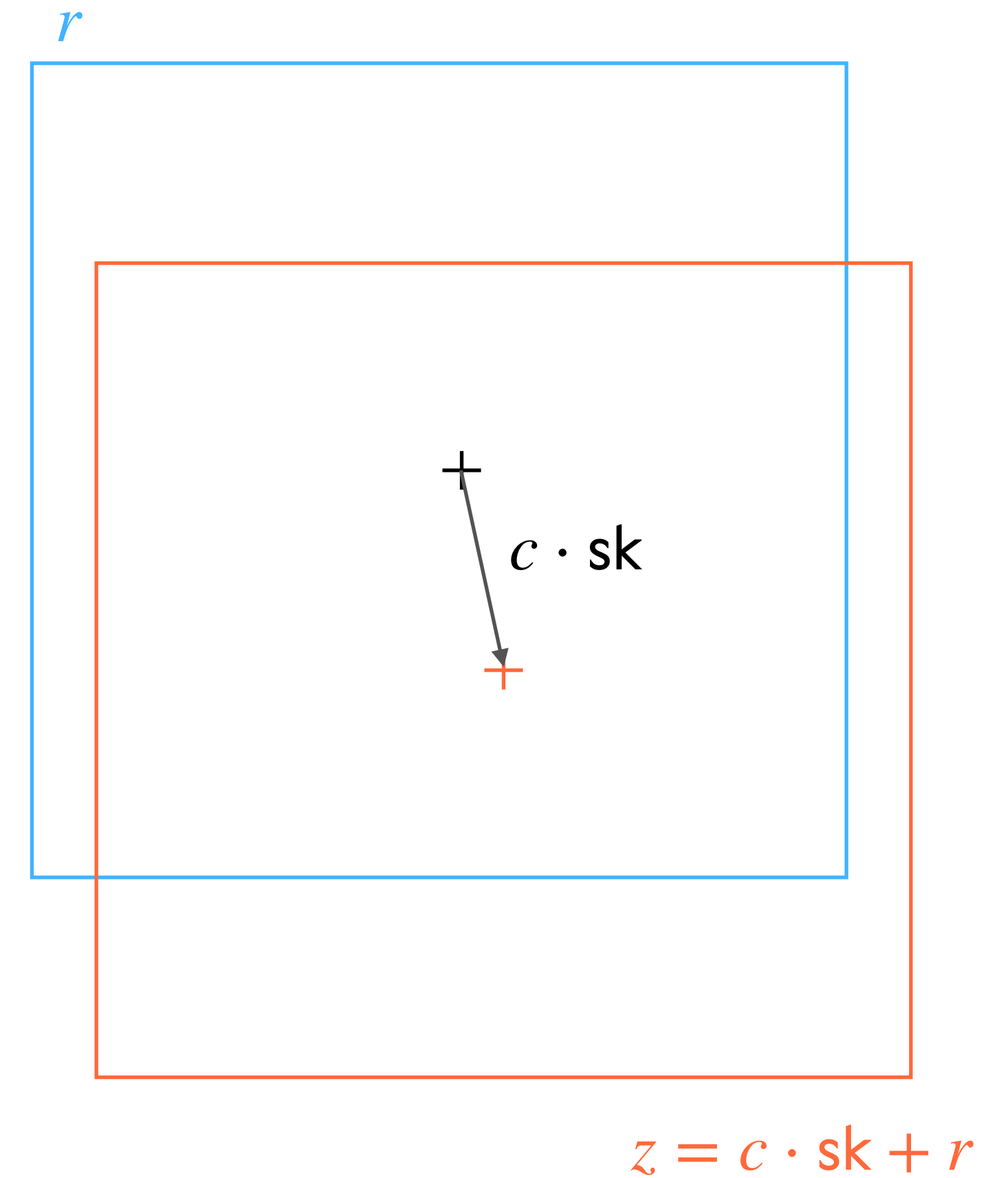
Sample r in a centered hypercube.



Rejection sampling

Sample r in a centered hypercube.

Then, the distribution of z depends on the secret.

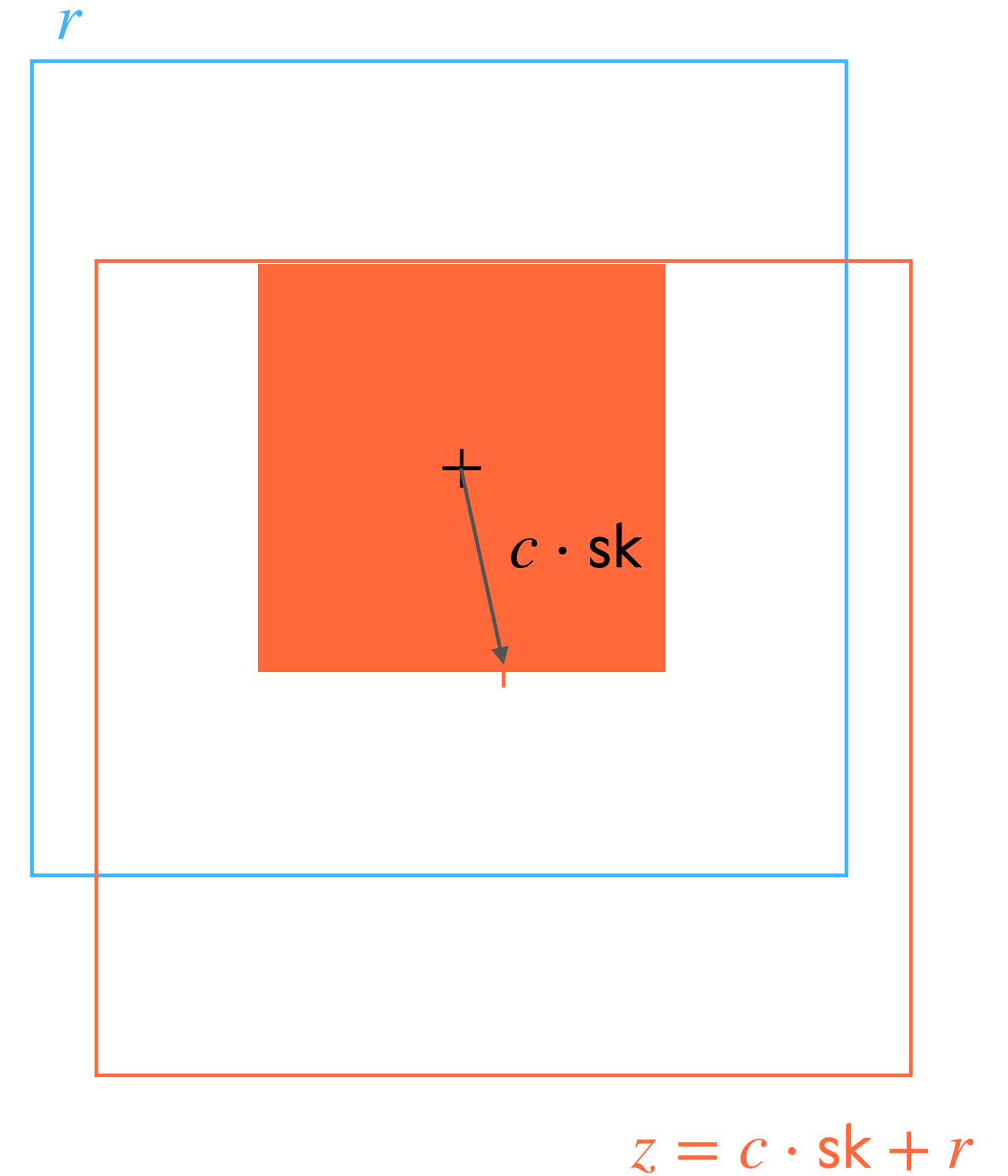


Rejection sampling

Sample r in a centered hypercube.

Then, the distribution of z depends on the secret.

We reject any z outside of .
The resulting distribution is independent of the secret.



ML-DSA signatures

ML-DSA . Keygen() $\rightarrow$ sk, vk

- $\text{vk} = \mathbf{A} \cdot \text{sk} + \mathbf{e}$, for sk, **e** short

ML-DSA . Sign(sk, msg) $\rightarrow$ sig

- Sample short **r**
- $\mathbf{w} = \lfloor \mathbf{A} \cdot \mathbf{r} \rfloor$
- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{z} = c \cdot \text{sk} + \mathbf{r}$
- If **z** not in S , **restart**
- If $\mathbf{A} \cdot \mathbf{z} - c \cdot \mathbf{e}$ not in S' , **restart**
- Output sig = (**z**, **w**)

MLWE assumption: vk appears uniformly distributed for **A** wide enough (more inputs than outputs)

ML-DSA . Verify(vk, msg, sig = (**z**, **w**))

- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{w} - (\mathbf{A} \cdot \mathbf{z} - c \cdot \text{vk})$ is short
- Assert **z** is small

ML-DSA signatures

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ML-DSA . Sign(sk, msg) $\rightarrow$ sig

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- $c = H(\mathbf{w}, \text{msg})$
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- If $\mathbf{z}$ not in S , **restart**
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- Output sig = ($\mathbf{z}$, $\mathbf{w}$)

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- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{w} - (\mathbf{A} \cdot \mathbf{z} - c \cdot \text{vk})$ is short
- Assert $\mathbf{z}$ is small

Short vector sampling and rejection sampling are hard to thresholdize

Threshold ML-DSA with generic MPC

ML-DSA . Keygen() $\rightarrow$ sk, vk

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First solution: Implement the signing algorithm with Generic MPC

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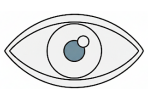
{firstname}.{lastname}@jpmchase.com

Threshold ML-DSA with generic MPC

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ML-DSA . Sign(sk, msg) $\rightarrow$ sig

- Sample short **r** * * * *
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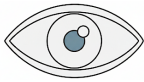
- Small modification in ML-DSA to safely reveal **w** and compute c in clear, relies on [dPN25]

Threshold ML-DSA with generic MPC

ML-DSA . Keygen() → sk, vk

- $vk = \mathbf{A} \cdot sk + \mathbf{e}$, for sk, **e** short

ML-DSA . Sign(sk, msg) → sig

- Sample short **r** * * * *
- $\mathbf{w} = \lfloor \mathbf{A} \cdot \mathbf{r} + \mathbf{e} \rfloor$ * * * *
- $c = H(\mathbf{w}, \text{msg})$ 
- $\mathbf{z} = c \cdot sk + \mathbf{r}$ * * * *
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- Output sig = (z, w)

First solution: Implement the signing algorithm with Generic MPC

Efficient, Scalable Threshold ML-DSA Signatures: An MPC Approach

Alexander Bienstock^{*‡}, Leo de Castro^{*†✉}, Daniel Escudero^{*‡}, Antigoni Polychroniadou^{*‡}, Akira Takahashi^{*‡}
JPMorgan Chase & Co.
^{*}JPMC AlgoCRYPT CoE, [†]JPMC CTC Cryptography, [‡]JPMC AI Research
`{firstname}.{lastname}@jpmchase.com`

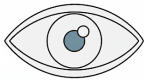
- Small modification in ML-DSA to safely reveal **w** and compute c in clear, relies on [dPN25]
- Batch comparisons for rejection sampling

Threshold ML-DSA with generic MPC

ML-DSA . Keygen() → sk, vk

- $vk = A \cdot sk + e$, for sk, e short

ML-DSA . Sign(sk, msg) → sig

- Sample short **r** * * * *
- $w = \lfloor A \cdot r + e \rfloor$ * * * *
- $c = H(w, msg)$  * * * *
- $z = c \cdot sk + r$ * * * *
- If **z** not in S , **restart** * * * *
- If $A \cdot z - c \cdot e$ not in S' , **restart** * * * *
- Output sig = (z, w)

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Concretely,

- Online:
 - 92 rounds w/ 1.2 MB comm
 - 24 rounds w/ 2.3 MB comm
- **Honest-majority** for better efficiency (can only tolerate $T/2$ corruptions for T signers)
- Offline: ?

Threshold ML-DSA for N parties ($T = N$)

ML-DSA^{*}.Keygen() $\rightarrow$ sk, vk

- For $1 \leq i \leq N$, $\text{vk}_i = \mathbf{A} \cdot \text{sk}_i + \mathbf{e}_i$, where sk, $\mathbf{e}_i$ short
- $\text{vk} = \sum_i \text{vk}_i$

Second solution: More tailored / using lattice properties
Sample N secrets, and aggregate the knowledge proofs.

ML-DSA . Verify(vk, msg, sig = ($\mathbf{z}$, $\lfloor \mathbf{w} \rfloor$))

- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- $\lfloor \mathbf{w} \rfloor - (\mathbf{A} \cdot \mathbf{z} - c \cdot \text{vk})$ is short
- Assert $\mathbf{z}$ is small

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ML-DSA^{*}.Sign(sk, msg) $\rightarrow$ sig

- For $1 \leq i \leq N$
 - Sample short $\mathbf{r}_i, \mathbf{e}'_i$
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Sample a $\mathbf{w}_i$ for each secret, and do not rely on rounding for security:
reintroduce error in $\mathbf{w}_i$ for rejection sampling on $\mathbf{e}$

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- For $1 \leq i \leq N$,
 - $\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}'_i$
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- For $1 \leq i \leq N$,

$$\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}'_i$$
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**We use more compact distributions
than ML-DSA to still pass verification
 $\rightarrow$ supports up to 6 parties**

Threshold ML-DSA for N parties ($T = N$)

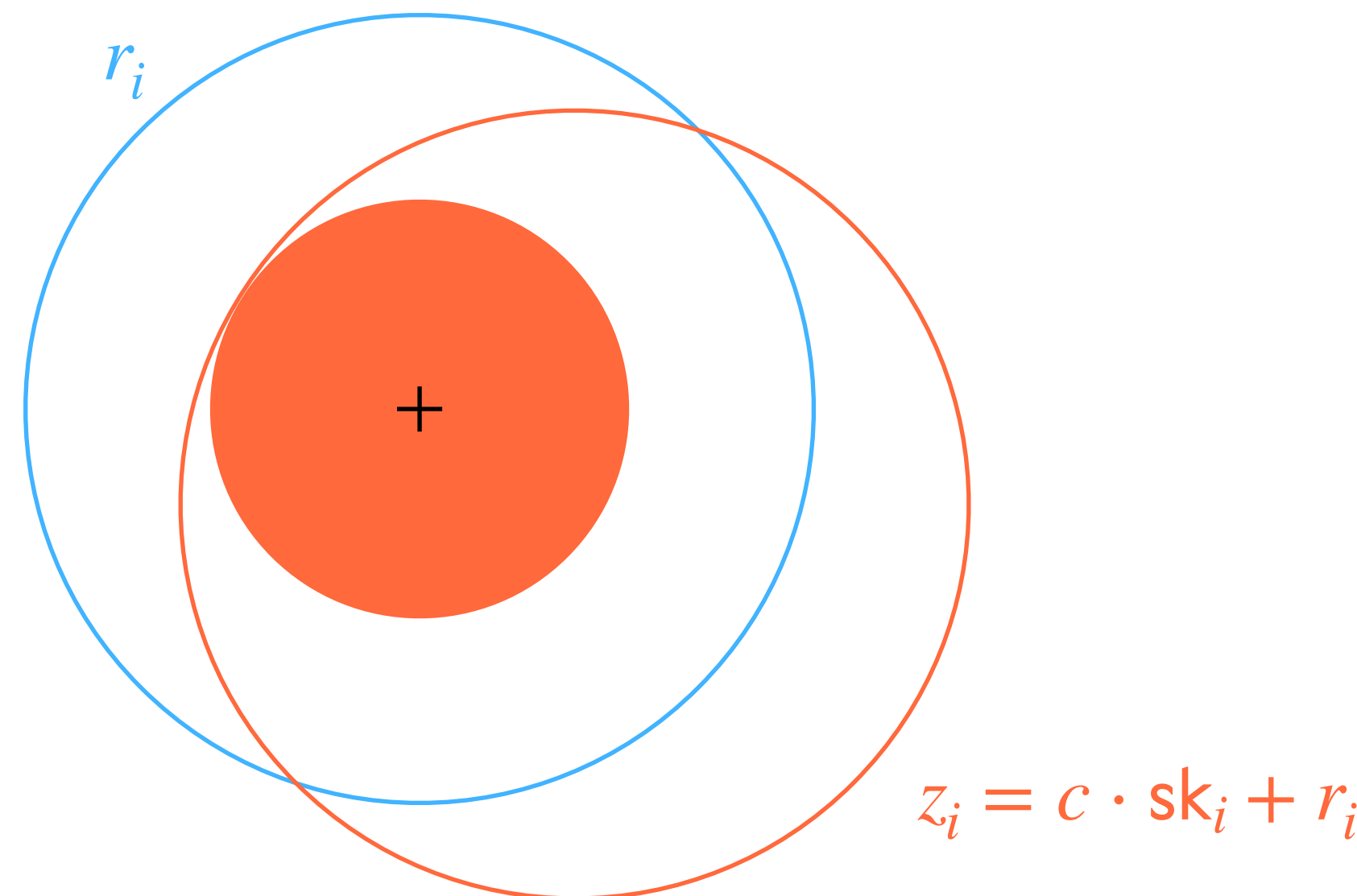
ML-DSA*.Keygen() →

- For $1 \leq i \leq N$, vk
- $\text{vk} = \sum_i \text{vk}_i$

ML-DSA*.Sign(sk, msg)

- For $1 \leq i \leq N$
 - Sample short
 - $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i +$
- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- For $1 \leq i \leq N$,
 - $\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}_i$
- If any $(\mathbf{z}_i, \mathbf{y}_i)$ not in S , **restart**
- $\text{sig} = (\sum_i \mathbf{z}_i, \lfloor \mathbf{w} \rfloor)$
- If sig not in S' , **restart**
- **return sig**

Rejection sampling with hyperballs



We use more compact distributions than ML-DSA to still pass verification
→ **supports up to 6 parties**

g lattice properties
e knowledge proofs.

Threshold ML-DSA for N parties ($T = N$)

ML-DSA^{*}.Keygen() $\rightarrow$ sk, vk

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ML-DSA^{*}.Sign(sk, msg) $\rightarrow$ sig

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 - Sample short $\mathbf{r}_i, \mathbf{e}'_i$
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- If sig not in S' , **restart**
- **return sig**

Th-ML-DSA . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample short $\mathbf{r}_i, \mathbf{e}'_i$
- Broadcast $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$

Round 2:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- $\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}'_i$
- If $(\mathbf{z}_i, \mathbf{y}_i)$ in S , **broadcast $\mathbf{z}_i$** , else **abort**

Combine:

- $\text{sig} = (\sum_i \mathbf{z}_i, \lfloor \mathbf{w} \rfloor)$
- If sig not in S' , **restart**
- **return sig**

Threshold ML-DSA for N parties ($T = N$)

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- $\text{sig} = (\sum_i \mathbf{z}_i, \lfloor \mathbf{w} \rfloor)$
- If sig not in S' , **restart**
- **return sig**

But, the scheme is only
secure if corrupted parties
do not bias $\mathbf{w}$

Th-ML-DSA . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample short $\mathbf{r}_i, \mathbf{e}'_i$
- Broadcast $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$

Round 2:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- $\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}'_i$
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- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{commit}_i = H(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$ + abort if inconsistent commit_i
- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- $\mathbf{z}_i = c \cdot \text{sk}_i + \mathbf{r}_i, \mathbf{y}_i = c \cdot \mathbf{e}_i + \mathbf{e}'_i$
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- $\text{sig} = (\sum_i \mathbf{z}_i, \lfloor \mathbf{w} \rfloor)$
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Threshold ML-DSA for $T \neq N$ parties

Use Replicated Secret Sharing [dPN25]

ML-DSA^{*}.Keygen() $\rightarrow$ sk, vk

- For every possible set I of $N - T + 1$ parties
 - $\text{vk}_I = \mathbf{A} \cdot \text{sk}_I + \mathbf{e}_I$, where $\text{sk}_I, \mathbf{e}_I$ short
 - Distribute $\text{sk}_I, \mathbf{e}_I$ to parties in I
- $\text{vk} = \sum_i \text{vk}_I$

1. When at most $T - 1$ parties are corrupted, at least one of these secrets remains hidden.

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1. When at most $T - 1$ parties are corrupted, at least one of these secrets remains hidden.
2. T parties can collaboratively reconstruct the full secret.

Partition $\sqcup_{i \in SS} m_i = \{I \text{ s.t. } |I| = N - T + 1\}$:

$$\text{sk} = \sum_{i \in SS} \sum_{I \in m_i} \text{sk}_I, \quad \mathbf{e} = \sum_{i \in SS} \sum_{I \in m_i} \mathbf{e}_I$$

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Th-ML-DSA . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{commit}_i = H(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$ + abort if inconsistent commit_i
- $c = H(\lfloor \mathbf{w} \rfloor, \text{msg})$
- $\mathbf{z}_i = c \cdot \sum_{I \in m_i} \text{sk}_I + \mathbf{r}_i, \mathbf{y}_i = c \cdot \sum_{I \in m_i} \mathbf{e}_I + \mathbf{e}'_i$
- If $(\mathbf{z}_i, \mathbf{y}_i)$ in S , broadcast $\mathbf{z}_i$, else abort

Combine:

- $\text{sig} = (\sum_i \mathbf{z}_i, \lfloor \mathbf{w} \rfloor)$
- If sig not in S' , restart
- return sig

Techniques from [dPN25].

Threshold ML-DSA for $T \neq N$ parties

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... plus some other optimizations to make parameters as tight as possible

Th-ML-DSA.Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$

broadcast $\text{commit}_i = H(\mathbf{w}_i)$

...

broadcast $\mathbf{w}_i$

...

$\sum_i \mathbf{w}_i$ + abort if inconsistent commit_i

$H(\lfloor \mathbf{w} \rfloor, \text{msg})$

$$c \cdot \sum_{I \in m_i} \text{sk}_I + \mathbf{r}_i, \mathbf{y}_i = c \cdot \sum_{I \in m_i} \mathbf{e}_I + \mathbf{e}'_i$$

- If $(\mathbf{z}_i, \mathbf{y}_i)$ in S , broadcast $\mathbf{z}_i$, else abort

Combine:

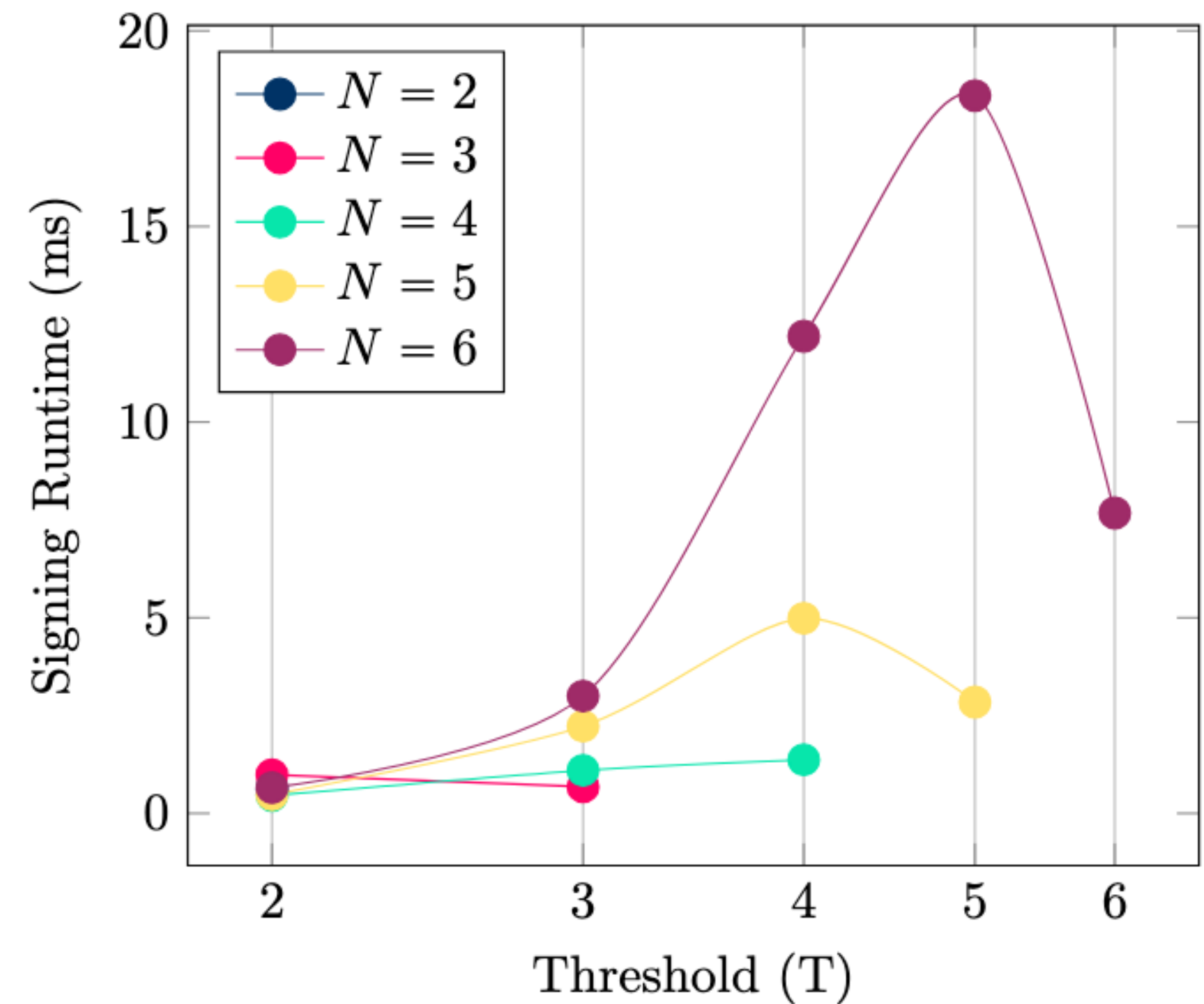
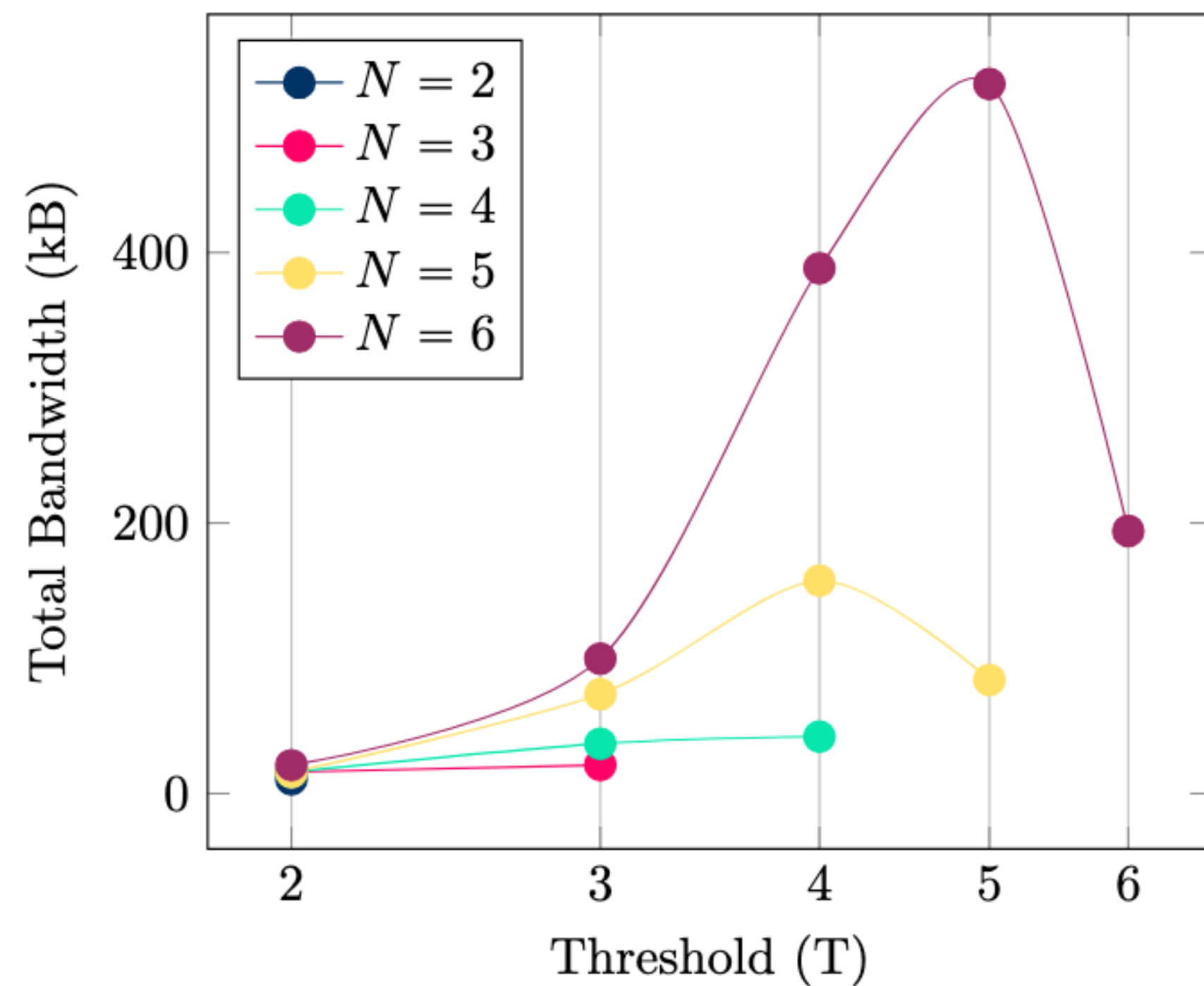
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- If sig not in S' , restart
- return sig

Techniques from [dPN25].

Evaluation

Parameters aim for a success probability $1/2$ for each attempt (vs $\sim 1/4$ in original ML-DSA).

Efficient up to 6 parties.



Bandwidth and latency of threshold signing for ML-DSA 44 (on a local network)

Threshold ML-DSA

Scheme	# Parties	# Rounds	Comm (MB)	Computation	Paradigm	Security
Our work	6	6	0.021 to 1.05	Lightweight	Game-based	Standard
Bienstock et al.	Unlimited	96	>1.2*	Online lightweight*	UC	Honest Majority
		24	>2.3*			

Average # rounds, and communication per party to obtain a valid signature

** Communication and computation exclude cost of offline correlated randomness generation.*

Threshold ML-DSA

Scheme	# Parties	# Rounds	Comm (MB)	Computation	Paradigm	Security
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Advanced properties?

DKG ✓

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		24	>2.3*			

Advanced properties?

DKG ✓

Identifiable Aborts ✗

Single online round ✗

Efficient for many parties ✗

Raccoon: ML-DSA without aborts

`Raccoon.Keygen()` $\rightarrow$ sk, vk

- $vk = \mathbf{A} \cdot sk + \mathbf{e}$, for $sk, \mathbf{e}$ short

`Raccoon.Sign( $sk, msg$ )` $\rightarrow$ sig

- Sample a short $\mathbf{r}, \mathbf{e}'$
- $\mathbf{w} = \mathbf{A} \cdot \mathbf{r} + \mathbf{e}'$
- $c = H(\mathbf{w}, msg)$
- $\mathbf{z} = c \cdot sk + \mathbf{r}$
- ~~If $\mathbf{z}$ not in S , restart~~
- Output $sig = (\mathbf{w}, \mathbf{z})$

Let's remove the rejection sampling!



Raccoon: ML-DSA without aborts

`Raccoon.Keygen()` $\rightarrow$ `sk, vk`

- $\text{vk} = \mathbf{A} \cdot \text{sk} + \mathbf{e}$, for `sk, e` short

`Raccoon.Sign(sk, msg)` $\rightarrow$ `sig`

- Sample a short $\mathbf{r}, \mathbf{e}'$
- $\mathbf{w} = \mathbf{A} \cdot \mathbf{r} + \mathbf{e}'$
- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{z} = c \cdot \text{sk} + \mathbf{r}$
- ~~If $\mathbf{z}$ not in S , restart~~
- Output `sig` = $(\mathbf{w}, \mathbf{z})$

Unforgeable under

- ♦ **Hint-MLWE**
- ♦ **SelfTargetMSIS**

vk	sig
2.3 kB	11.5 kB

Hint-MLWE assumption [KLSS23].

$(\mathbf{A}, \text{vk})$ is pseudorandom even if given
 Q “hints”:

$$(c_i, \mathbf{z}_i := c_i \cdot \text{sk} + \mathbf{r}_i) \text{ for } i \in [Q]$$

As hard as MLWE_σ if

$$\sigma_r \geq \sqrt{Q} \cdot \|c\| \cdot \sigma$$

Threshold Raccoon

Raccoon.Keygen() $\rightarrow$ sk, vk

- $\text{vk} = \mathbf{A} \cdot \text{sk} + \mathbf{e}$, for sk, **e** short

Raccoon.Sign(sk, msg) $\rightarrow$ sig

- Sample a short **r**, **e'**
- $\mathbf{w} = \mathbf{A} \cdot \mathbf{r} + \mathbf{e}'$
- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{z} = c \cdot \text{sk} + \mathbf{r}$
- Output sig = (**w**, **z**)

Raccoon.Verify(vk, msg, sig = (w**, **z**))**

- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{y} = \mathbf{w} - \mathbf{A} \cdot \mathbf{z} + c \cdot \text{vk}$
- Assert (**y**, **z**) short

Shamir sharing on secret $\text{sk} \in \mathcal{R}_q^\ell$

Sample polynomial $f \in \mathcal{R}_q^\ell[X]$ s.t.

- $f(0) = \text{sk}$ and $\deg f \leq T - 1$
- Partial signing keys $\text{sk}_i := \llbracket \text{sk} \rrbracket_i = f(i)$

Properties:

- with $< T$ shares, sk is perfectly hidden
- with a set S of $\geq T$ shares, reconstruct sk via Lagrange interpolation

$$\text{sk} = \sum_{i \in S} L_{S,i} \cdot \llbracket \text{sk} \rrbracket_i$$

Threshold Raccoon

Raccoon.Keygen() $\rightarrow$ sk, vk

- $\text{vk} = \mathbf{A} \cdot \text{sk} + \mathbf{e}$, for sk, **e** short

Raccoon.Sign(sk, msg) $\rightarrow$ sig

- Sample a short **r**, **e'**
- $\mathbf{w} = \mathbf{A} \cdot \mathbf{r} + \mathbf{e}'$
- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{z} = c \cdot \text{sk} + \mathbf{r}$
- Output sig = (**w**, **z**)

Raccoon.Verify(vk, msg, sig = (w**, **z**))**

- $c = H(\mathbf{w}, \text{msg})$
- $\mathbf{y} = \mathbf{w} - \mathbf{A} \cdot \mathbf{z} + c \cdot \text{vk}$
- Assert (**y**, **z**) short

First (insecure) attempt

ThRaccoon.Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short **r_i**, **e'_i**
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast **w_i**

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket \text{sk} \rrbracket_i + \mathbf{r}_i$

Combine: the final signature is

$$(\mathbf{w}, \sum_{i \in S} \mathbf{z}_i)$$

Threshold Raccoon

- ◆ Prevent ROS attack with commit-reveal of $\mathbf{w}_i$

First (insecure) attempt

ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket sk \rrbracket_i + \mathbf{r}_i$

Combine: the final signature is

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Threshold Raccoon

- ◆ Prevent ROS attack with commit-reveal of $\mathbf{w}_i$
- ◆ But, $\mathbf{r}_i$ is small vs $L_{S,i} \cdot c \cdot \llbracket sk \rrbracket_i$ is large
→ Leaks $\llbracket sk \rrbracket_i$

First (insecure) attempt

ThRaccoon . Sign(sk, msg) → sig

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Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
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Threshold Raccoon

- ◆ Prevent ROS attack with commit-reveal of $\mathbf{w}_i$
- ◆ But, $\mathbf{r}_i$ is small vs $L_{S,i} \cdot c \cdot \llbracket sk \rrbracket_i$ is large
→ Leaks $\llbracket sk \rrbracket_i$
- ◆ Solution: add a zero-share Δ_i :
 - Derived with a PRF, using pre-shared pairwise keys
 - Any set of $< T$ values Δ_i is uniformly random
 - $\sum_{i \in S} \Delta_i = 0$

ThRaccoon . Sign(sk, msg) → sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket sk \rrbracket_i + \mathbf{r}_i + \Delta_i$

Combine: the final signature is

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Threshold Raccoon

max N	Speed	Rounds	vk	sig	Total communication
1024	Fast	3	4 kB	13 kB	40 kB

Two-round ThRaccoon [EKT24]

2Rnd-ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample short $\mathbf{r}_{i,j}, \mathbf{e}'_{i,j}$ for $j \in [\text{rep}]$
- $\mathbf{w}_{i,j} = \mathbf{A} \cdot \mathbf{r}_{i,j} + \mathbf{e}'_{i,j}$ for $j \in [\text{rep}]$
- Broadcast $(\mathbf{w}_{i,j})_j$

Round 2:

- Derive coefficients $(\beta_{i,j})_{i,j} = H((\mathbf{w}_{i,j})_{i,j})$
- $\mathbf{w} = \sum_{i,j} \beta_{i,j} \mathbf{w}_{i,j}$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket \text{sk} \rrbracket_i + \sum_j \beta_{i,j} \mathbf{r}_{i,j} + \Delta_i$

Combine: the final signature is

$$(\mathbf{w}, \sum_{i \in S} \mathbf{z}_i)$$

ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket \text{sk} \rrbracket_i + \mathbf{r}_i + \Delta_i$

Combine: the final signature is

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2Rnd-ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

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- $\mathbf{w}_{i,j} = \mathbf{A} \cdot \mathbf{r}_{i,j} + \mathbf{e}'_{i,j}$ for $j \in [\text{rep}]$
- Broadcast $(\mathbf{w}_{i,j})_j$

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Combine: the final signature is

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Unforgeable under

- ♦ **AOMLWE**
- ♦ **SelfTargetMSIS**

vk	sig
5.5 kB	10.8 kB

~270 kB communication (offline + online): 5x TRaccoon

AOMLWE assumption.

$(\mathbf{A}, \text{vk})$ is pseudorandom even if given many $(\mathbf{w}_j)$, and adaptively choosing coefficients β_j to obtain:

$$(c_i, \mathbf{z}_i := c_i \cdot \text{sk} + \sum_j \beta_j \cdot \mathbf{r}_j) \text{ for } i \in [Q]$$

Detecting signing failures origins

Identify aborts in ThRaccoon with NIZK

- We can identify aborts using NIZK.
- Hard to prove to prove correct computation of Δ_i , but we can prove all other computations with the NIZK
- At a high-level, $\Delta_i = \sum_j m_{i,j}$ where $m_{i,j}$ is the output of a PRF known by i and j .
 - ◆ We can simply check that all i, j agree on $m_{i,j}$, if not one cheated and we can reveal the corresponding PRF key to check computations.

NIZK-ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = L_{S,i} \cdot c \cdot \llbracket sk \rrbracket_i + \mathbf{r}_i + \Delta_i$

Combine: the final signature is

$$(\mathbf{w}, \sum_{i \in S} \mathbf{z}_i)$$

DKG + Detecting signing failures origins

Identify aborts in ThRaccoon with sDKG

- We can also leverage a new class of secret sharings: “short secret sharing”
 - ◆ Secret sharing with shares of small norms: partial leak, but ok for lattices
 - ◆ In this case, we can remove the zero-share Δ_i !

sDKG-ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

Round 2:

- Broadcast $\mathbf{w}_i$

Round 3:

- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = c \cdot \langle L_{S,i} \cdot \text{sk}_i \rangle + \mathbf{r}_i$

Combine: the final signature is

$$(\mathbf{w}, \sum_{i \in S} \mathbf{z}_i)$$

DKG + Detecting signing failures origins

Identify aborts in ThRaccoon with sDKG

- We can also leverage a new class of secret sharings: “short secret sharing”
 - ◆ Secret sharing with shares of small norms: partial leak, but ok for lattices
 - ◆ In this case, we can remove the zero-share Δ_i !
- Identifiable abort for free, but larger private key
- Scales up to 64 parties

sDKG-ThRaccoon . Sign(sk, msg) $\rightarrow$ sig

Round 1:

- Sample a short $\mathbf{r}_i, \mathbf{e}'_i$
- $\mathbf{w}_i = \mathbf{A} \cdot \mathbf{r}_i + \mathbf{e}'_i$
- Broadcast $\text{cmt}_i = H_{\text{cmt}}(\mathbf{w}_i)$

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- $\mathbf{w} = \sum_i \mathbf{w}_i$
- $c = H(\mathbf{w}, \text{msg})$
- Broadcast $\mathbf{z}_i = c \cdot \langle L_{S,i} \cdot \text{sk}_i \rangle + \mathbf{r}_i$

Combine: the final signature is

$$(\mathbf{w}, \sum_{i \in S} \mathbf{z}_i)$$

Conclusion

Conclusion

- Diverse proposal for Threshold Signatures: lattices, multivariate, hash-based, isogenies
- Practical schemes compatible with **ML-DSA, UOV and MAYO**
- Possible fallback to Threshold Raccoon for large thresholds / advanced properties

Questions?



Evaluation

Other ML-DSA parameter sets

