Polytopes in the Fiat-Shamir with Aborts Paradigm CWI Student Seminar

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• Talk based on https://eprint.iacr.org/2024/411.pdf.

I. Intro: Fiat-Shamir and Rejection Sampling

II. The Polytope-based Framework

III. Choosing a Polytope \mathcal{H}

IV. Sampling in $\mathcal{H} \cap \mathbb{Z}^n$

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SIS-based ID protocol



Fiat-Shamir with Aborts



Rejection sampling: a brief history of distributions

Idea: provably transform an imperfect distribution into a perfect distribution.



Our security relies on structured variants of SIS: MLWE, MSIS and SelfTargetMSIS.

The important metric for signature size and $Supp(V_{cs})$ is the L_2 metric.

- We focus on the unimodal case (for now).
- We focus on uniform distributions.
- Notation: we identify distribution V_y and set $Supp(V_y)$.

Rejection sampling: motivation



Knowing z should reveal no information on y and cs.

Rejection sampling: motivation



Witness-Indistinguishability: each z in the blue area is equally likely to have been generated from any valid secret key.

Rejection sampling: motivation



Witness-Indistinguishability: each z in the blue area is equally likely to have been generated from any valid secret key.

This must hold for **all** elements of V_{cs} .

Assuming uniform distributions z avoids information leakage if and only if:

$$V_z \subseteq \bigcap_{\mathbf{x}\in V_{cs}} (V_y + \mathbf{x}).$$

Furthermore, V_{τ} minimises the number of rejects if and only if:

$$V_z = \bigcap_{\mathbf{x} \in V_{cs}} (V_y + \mathbf{x}).$$

- max_{z∈Vz} ||z||₂ conditions the signature size.
 Rejection rate depends on the tightness of the inclusion.

Illustration: a Square



Illustration: a Square



Illustration: a Square



Probability of rejecting:
$$\frac{\text{Vol}(V_z)}{\text{Vol}(V_y)}.$$

 [DFPS22] observe that Gaussian distributions and uniform distributions in Hyperballs give optimal sizes.





- Very small sizes (optimal according to [DFPS22]).
- Hard to mask against side channels.
- Hard to sample (Fixed point arithmetic).
- Only analysed in the continuous setting.
- Used in HAETAE [CCD⁺23].

- Larger sizes (in some sense hard to do worse).
- Easy to mask against side channels.
- Very simple sampler.
- Valid in the discrete setting.
- Used in DILITHIUM [DKL+21].



What we want:

- Good proof sizes (better than DILITHIUM).
- A simple sampler (no FP arithmetic and no Gaussians).
- A valid analysis in the discrete setting.

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Definition (Polytope)

A *polytope* is the convex hull of its vertices $\mathcal{V}(\mathcal{P}) = {\mathbf{x}_1, \dots, \mathbf{x}_v} \in \mathbb{R}^n$.



Theorem (\mathcal{P} -ception: Intersection of polytopes)

Let \mathcal{P} be a symmetric inscriptible and circumscriptible polytope. Let $r, R \in \mathbb{R}_{>0}$ such that R > r. Then:

$$\bigcap_{\mathbf{c}\in\mathcal{P}_r}\mathcal{P}_{R,\mathbf{c}}=\bigcap_{\mathbf{c}\in\mathcal{V}(\mathcal{P}_r)}\mathcal{P}_{R,\mathbf{c}}=\bigcap_{one\ \mathbf{c}_i\ per\ face\ of\ \mathcal{P}_r}\mathcal{P}_{R,\mathbf{c}_i}=\mathcal{P}_{R-r}$$

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Corollary (Discrete version)

If $\mathcal{V}(\mathcal{P}_r) \subset \mathbb{Z}^n$, then

$$\bigcap_{\mathbf{c}\in\mathcal{P}_{r,\mathbb{Z}}}\mathcal{P}_{R,\mathbf{c}}=\bigcap_{\mathbf{c}\in\mathcal{V}(\mathcal{P}_r)}\mathcal{P}_{R,\mathbf{c},\mathbb{Z}}=\mathcal{P}_{R-r,\mathbb{Z}},$$

where $\mathcal{P}_{\mathbb{Z}} = \mathcal{P} \cap \mathbb{Z}^n$.

\mathcal{P} -ception: Illustration 1





\mathcal{P} -ception: Illustration 2



Rejection Sampling with Polytopes: Continuous case

Let \mathcal{P}^n be a symmetric polytope whose vertices all lie on a sphere.

Theorem (informal)

If $V_y = \mathcal{P}_R^n$ and $V_{cs} \subseteq \mathcal{P}_r^n$, then:

$$\frac{\operatorname{Vol}\mathcal{P}_{R-r}^n}{\operatorname{Vol}\mathcal{P}_{R}^n} = \left(\frac{R-r}{R}\right)^n$$

determines the rejection rate.

In practical instantiations, $r \ll R$.

Rejection Sampling with Polytopes: Discrete case

Let \mathcal{P}^n be a symmetric polytope, with integral vertices all on a sphere, then:

Theorem (informal)

If $V_v = \mathcal{P}_R^n \cap \mathbb{Z}^n$ and $V_{cs} \subseteq \mathcal{P}_r^n \cap \mathbb{Z}^n$, then:

$$\frac{|\mathcal{P}_{R-r,\mathbb{Z}}^{n}|}{|\mathcal{P}_{R,\mathbb{Z}}^{n}|} = \frac{\operatorname{Vol}\mathcal{P}_{R-r}^{n}}{\operatorname{Vol}\mathcal{P}_{R}^{n}} \cdot \frac{|\mathcal{P}_{R-r,\mathbb{Z}}^{n}|}{\operatorname{Vol}\mathcal{P}_{R-r}^{n}} \cdot \frac{\operatorname{Vol}\mathcal{P}_{R}^{n}}{|\mathcal{P}_{R-r,\mathbb{Z}}^{n}|} = \left(\frac{R-r}{R}\right)^{n} \frac{1+\varepsilon_{R}}{1+\varepsilon_{R-r}}$$

determines the rejection rate.

Computing ε should be done only once, and requires:

- Volumes of integral polytopes.Counting integral points in polytopes.

Efficient for well-chosen polytopes

Extra motivation: Optimality of rejection



Recall that we would like maximality of:

 $\bigcap_{\mathbf{x}\in V_{cs}}(V_y+\mathbf{x}).$

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Recall that we would like maximality of:

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If the support V_y is a polytope, and if \mathcal{P} is a symmetric polytope that admits an inscribed ball \mathcal{B}_2 that is tangent to all of its faces, then we can interchangeably use \mathcal{P} or \mathcal{B}_2 for the support of **cs**.

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What we want for $\mathcal{P}:$

- . Symmetric
- . Inscriptible
- . Circumscriptible
- . Small ratio
- . Integral vertices
- . Efficiently samplable



Interlude: High-dimensional balls



The Hypercube: $\mathcal{B}_{\infty}(R) = \{ \mathbf{x} \in \mathbb{R}^n : orall i, |x_i| \leq R \}.$

- Norm: L_{∞} .
- Volume: $(2R)^n$.
- Inradius: R.
- Circumradius: \sqrt{nR} .
- Mass concentrates: at the corners.

The Cross-polytope¹:
$$\mathcal{B}_1(R) = \{ \mathbf{x} \in \mathbb{R}^n : \sum |x_i| \leq R \}.$$

- Norm: L_1 .
- Volume: $\frac{(2R)^n}{n!}$.
- Inradius: $\frac{1}{\sqrt{n}}R$.
- Circumradius: R.
- Mass concentrates: at the center.



¹also called Hyperoctahedron, Orthoplex, or Cocube.

$$\mathcal{H}_r^n = \mathcal{B}_\infty^n(r) \cap \mathcal{B}_1^n(r\sqrt{n})$$







Some properties of ${\mathcal H}$

• Volume \approx Vol($\mathcal{B}_1^n(r\sqrt{n}))$:

$$\frac{(2r\sqrt{n})^n}{n!}\sum_{i=0}^{\lfloor\sqrt{n}\rfloor}(-1)^i\binom{n}{i}\left(1-\frac{i}{\sqrt{n}}\right)^{n+1}$$

- Inradius: r (by design).
- Circumradius:

$$r\sqrt{\lfloor\sqrt{n}
floor}+(\sqrt{n}-\lfloor\sqrt{n}
floor)^2\leq r\sqrt[4]{n}.$$

 ${\cal H}$ is symmetric, and perfectly inscriptible and circumscriptible.



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The following sets are isomorphic via a simple projection:

$$S_{1,\mathbb{Z}^+}^{n+1}(r\sqrt{n}) = \{ \mathbf{y} \in \mathbb{Z}_{\geq 0}^{n+1} : \|\mathbf{y}\|_1 = r\sqrt{n} \},\$$
$$B_{1,\mathbb{Z}^+}^n(r\sqrt{n}) = \{ \mathbf{y} \in \mathbb{Z}_{\geq 0}^n : \|\mathbf{y}\|_1 \le r\sqrt{n} \}.$$

Bonus trick: project away from the largest coordinate to lower $\mathbb{E}(\|\mathbf{y}\|_{\infty})$.



Sampling in $\mathcal{H} \cap \mathbb{Z}^n$

SampleL1Sphere(n, r)		SampleL1Ball(n, r)		
1:	$/\!\!/ \ S = \{X \subset [\![1, r + n - 1]\!] : \#X = n - 1\}$	1:	$(y_i)_{n+1} \stackrel{\$}{\leftarrow} \texttt{SampleL1Sphere}(n+1,r)$	
2:	$X \stackrel{\hspace{0.1em}\scriptscriptstyle\$}{\leftarrow} \mathcal{U}(\mathcal{S})$	2:	return (y_1, \cdots, y_n)	
3:	$\mathbf{X} \leftarrow \{0\} \cup \mathbf{X} \cup \{r+n\}$	-		
4:	X.sort()	SampleH(n,r)		
5:	$/\!\!/ x_0, \cdots, x_n$ the ordered elements of X	1:	$\Delta_n \leftarrow (\sqrt{n} - \lfloor \sqrt{n} floor)$	
6:	for $i \in \llbracket 1, n rbracket$:	2:	$r' \leftarrow \lfloor \sqrt{n} floor r + \lfloor \Delta_n r floor$	
7:	$b \stackrel{\$}{\leftarrow} \{0,1\}$	3:	$\mathbf{Y} \leftarrow \bot$	
8:	$y_i \leftarrow (x_i - x_{i-1} - 1)$	4:	while $\mathbf{Y}=\perp$ do	
9:	if $y_i + b = 0$ then	5:	$\mathbf{Y} \gets \texttt{SampleL1Ball}(n,r')$	
10 :	restart	6:	if $\ \mathbf{Y}\ _{\infty} > r$ then	
11 :	$y_i \leftarrow (-1)^b y_i$	7:	$\mathbf{Y} \leftarrow ot$	
12 :	return $\mathbf{Y} := (y_i)_{1 \le i \le n}$	8:	return Y	

Mind the sides!

- Flip *n* coins for signs.
- Restart for each 0 coordinate, with probability 1/2.
- . Uniform: 🗸
- . Isochronous: 🗸
- . Expected restarts: small if $n \ll r$.



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 - $n \ll r$.



We have simple sampling with quality $n^{1/4}$

$$\mathcal{C}^n_{ heta,r} = \mathcal{H}^n_r \cap \mathcal{B}_2(heta \cdot r)$$
 where $heta pprox 1.5$

$$\label{eq:keyobservation:for} \left[\begin{array}{c} {\sf Key \ observation: \ for \ } \theta > c, \\ 1 - \exp(-\sqrt{n}) < \frac{{\sf Vol} \ {\cal C}^n_{\theta,r}}{{\sf Vol} \ {\cal H}^n_r} < 1. \end{array} \right]$$

- Ratio $n^{1/4} \rightarrow \theta$
- Trade-off between aborts and size.
- Warning: not a polytope anymore.

A new Fiat-Shamir with Aborts signature scheme: PATRONUS



Signature performances: Concrete (example) parameters

- Signature sizes: (in bytes)

Security target (bits)	120	180	260
HAETAE	1,463	2,337	2,908
PATRONUS ² (this work)	1,869	2,398	3,459
DILITHIUM	2,420	3,293	4,595

- Verification key sizes: Similar to DILITHIUM \checkmark
- Expected rejects: Similar to HAETAE \checkmark
- Sampler randomness: at most 1.3 times that of DILITHIUM \checkmark
- Optimised sampler implementation: Work in progress 🛦

²Parameters may still vary

What you should remember:

- We propose a new framework for rejection sampling in polytopes.
- This allows for rigorous analysis of perfect rejection in Fiat-Shamir.
- Our polytope $\mathcal H$ uses L_1 and L_∞ balls to approach an optimal L_2 ball.
- It is easy to sample from $\mathcal{H}_{\mathbb{Z}}$.
- This leads to the signature scheme PATRONUS, an interesting tradeoff between DILITHIUM and HAETAE.

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V. Bonus: Open Questions and Perspectives

Can we get a better polytope?

Theorem (From [Kas77])

There exists a constant 1 < c < 32 such that for each n, there exists an orthogonal $U \in O_n(\mathbb{R})$ such that

 $\mathcal{B}_2^n(1)\subseteq \mathcal{B}_1^n(\sqrt{n})\cap U\mathcal{B}_1^n(\sqrt{n})\subseteq \mathcal{B}_2^n(c).$

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Objective: Use the trick by [DDLL13] for better sizes.

- We need to study

$$I = igcap_{\mathbf{sc}\in\mathcal{B}_2(r)} \left(\mathcal{P}_{R,\mathbf{sc}}\cup\mathcal{P}_{R,-\mathbf{sc}}
ight)$$

- No improvement in the Hypercube case.
- For \mathcal{H} , no obvious improvement after dim 4 as the largest \mathcal{H} in I is \mathcal{H}_{R-r} .
- For $\ensuremath{\mathcal{C}}$, less unlikely.



Thank you for listening!



If you have extra questions, feel free to contact Hugo (hugo.beguinet@ens.fr)

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