

Rounded Gaussians

Fast and Secure Constant-Time Sampling for Lattice-Based
Crypto

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Lattice-based signatures

- ▶ Bimodal Lattice Signature Scheme (BLISS) (CRYPTO '13 by Léo Ducas and Alain Durmus and Tancreède Lepoint and Vadim Lyubashevsky)
- ▶ Pretty short and efficient; already included in [strongSwan](#) (library for IPsec-based VPN).
- ▶ Needs noise from discrete Gaussian distribution.
- ▶ CHES 2016: [Flush, Gauss, and Reload A Cache-Attack on the BLISS Lattice-Based Signature Scheme](#) by Groot Bruinderink, Hülsing, Lange, and Yarom.
- ▶ ACM-CCS 2017: [To BLISS-B or not to be – Attacking strongSwan's Implementation of Post-Quantum Signature](#) by Pessl, Groot Bruinderink, and Yarom.

Simplified BLISS

- ▶ Work in $R = \mathbf{Z}[x]/(x^n + 1)$, $n = 2^r$, and $R_q = (\mathbf{Z}/q)[x]/(x^n + 1)$ for q prime.
- ▶ Secret key $S = (s_1, s_2) = (f, 2g + 1) \in R_q^2$, f, g sparse in $\{0, \pm 1\}^n$.
- ▶ Public key $A = (a_1, a_2) \in R_{2q}^2$, with key equation $a_1 s_1 + a_2 s_2 \equiv q \pmod{2q}$.
- ▶ Computed as $a_q = (2g + 1)/f \pmod{q}$ (restart if f is not invertible); then $A = (2a_q, q - 2) \pmod{2q}$.

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- ▶ Can verify key guess for f with key equation; g computable.
- ▶ To sign, sample y from discrete n -dim Gaussian $D_{\mathbf{Z}^n, \sigma}$.
- ▶ $c = H(a_1, y, \text{public stuff})$ // H special hash function.
- ▶ choose a random bit b .
- ▶ Signature: (z, c) with $z = y + (-1)^b s_1 \cdot c \pmod{2q}$.
- ▶ Can get $\pm s_1 = (z - y)/c \in R_q$ if we know y , the error vector/polynomial; (c needs to be invertible).

Discrete Gaussian

- ▶ Continuous Gaussian

$$\rho_{\nu, \sigma}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(\frac{-\|s - \nu\|^2}{2\sigma^2}\right)$$

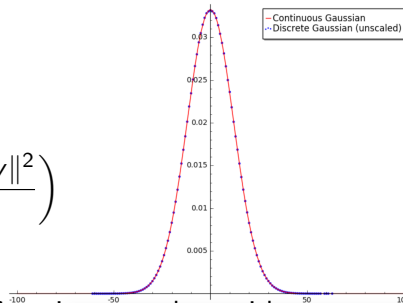
with center ν and variance s .

- ▶ Take integer values of continuous Gaussian; sample x with probability

$$D_{\nu, \sigma}(x) = \rho_{\nu, \sigma}(x) / \rho_{\sigma}(\mathbf{Z}),$$

where $\rho_{\sigma}(\mathbf{Z}) = \sum_{z \in \mathbf{Z}} \rho_{\sigma}(z)$.

- ▶ Complicated to do in practice; relatively nice to analyze.
- ▶ Do this m times for m -dimensional discrete Gaussian.



Rounded Gaussian

- ▶ Idea: Sample $z \in \mathcal{R}$ from continuous Gaussian, round to nearest integer $x \in \mathbf{Z}$; output x .

Rounded Gaussian

- ▶ Idea: Sample $z \in R$ from continuous Gaussian, round to nearest integer $x \in \mathbf{Z}$; output x .

- ▶ In math:

$$R_{\nu,\sigma}(x) = \int_{x-0.5}^{x+0.5} \rho_{\nu,\sigma}(s) ds.$$

- ▶ Easy to implement, harder to analyze.
- ▶ Hard part: needed to redo all proofs for BLISS etc..
- ▶ Box-Muller sampling is <50 lines of code on top of Fog's VCL (constant-time vectorized implementation of sin, cos, log, sqrt, ...)