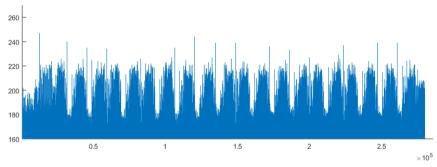
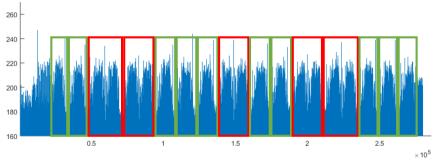


More Practical Single-Trace Attacks on the Number Theoretic Transform

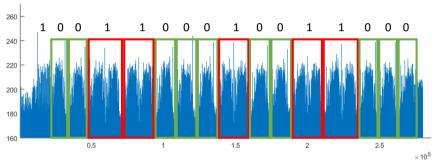
Peter Pessl, Robert Primas Graz University of Technology LATINCRYPT 2019. October 02



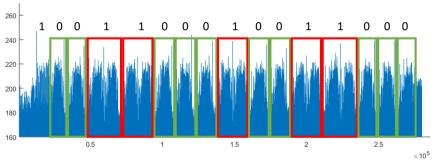
Power consumption trace of RSA decryption



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Single-trace attacks are still a prime threat!

But RSA is old news anyway...

- Lattice-based cryptography
 - promising post-quantum replacement
 - implementations: fast and constant time / control flow
- Do we still need to worry about single-trace attacks'
 - no more instruction leakage
 - protection efforts towards differential (multi-trace) attacks

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- Our previous work: single-trace attack on the NTT
 - Number Theoretic Transform, common in many lattice schemes
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Can we do better?

Our Contribution

- Improve upon previous attack
 - several improvements to belief propagation in this context
 - change targets: encryption instead of decryption
- Attack constant-time ASM-optimized Kyber implementation
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"Noisy ElGamal" with polynomials in $\mathbb{Z}_q[x]/\langle x^n+1\rangle$

Key Generation: generate small error polynomials s, e

$$t = a \cdot s + e$$

$$pk = (a, t), sk = s$$

Encryption: generate small error polynomials r, e_1, e_2

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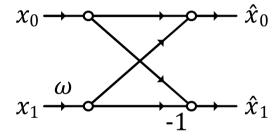
Number Theoretic Transform

- Naive polynomial multiplication: $\mathcal{O}(n^2)$
- Better: Number Theoretic Transform (NTT)
 - \approx FFT in $\mathbb{Z}_q[x]$, runtime $\mathcal{O}(n \log n)$
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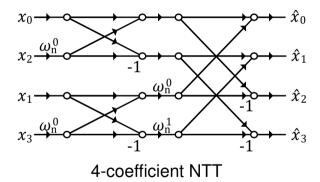
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Butterfly



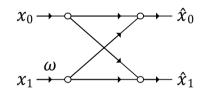
Butterfly = 2-coefficient NTT

Butterfly Network



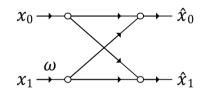
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- Profile power consumption of mult
- Match profiles (templates) for probability distribution



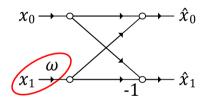
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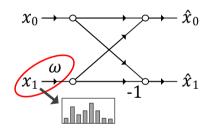
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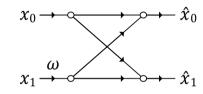
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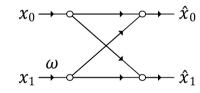
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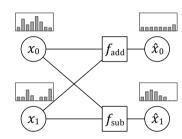
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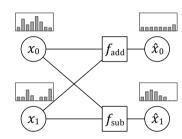
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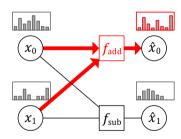
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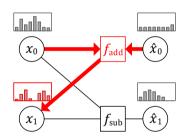
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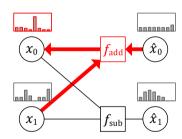
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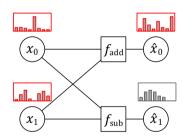
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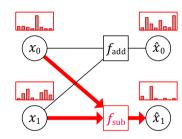
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- Evaluation on non-constant-time implementation
 - timing information not needed per se
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- Requires powerful attacker
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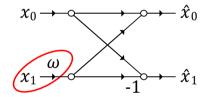
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Decreased Templating Effort

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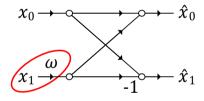
Previously



Target multiplication
1 million multivariate templates

Decreased Templating Effort

Previously



Target multiplication

1 million multivariate templates

Now \hat{x}_0 \hat{x}_0 \hat{x}_0 \hat{x}_1

Target memory loads and stores 14 univariate Hamming-weight templates

Are we done?

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Decryption

$$m \approx c_2 - \text{INTI}(\text{NTI}(s) \circ \text{NTI}(c_1))$$

Recover INTT input, compute s
INTT input: $[0, q - 1]^n$

Encryption

$$c_1 = \text{INTT}(\text{NTT}(a) \circ \text{NTT}(r)) + e_1$$

Recover r , compute $m \approx c_2 - t \cdot r$
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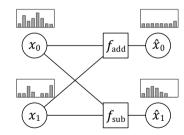
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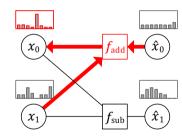
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Attack simulations already work, but we can do better...

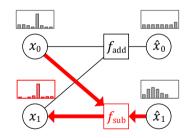
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 - overconfidence, non-covergence
 - short loop, deterministic operations



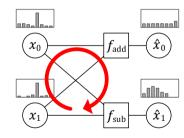
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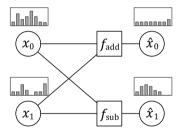
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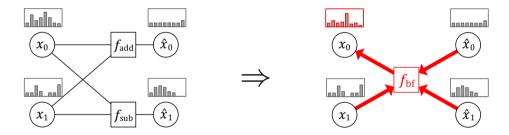
Butterfly Factors



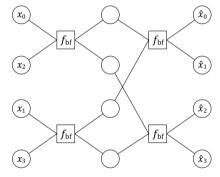
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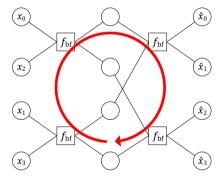


Still...



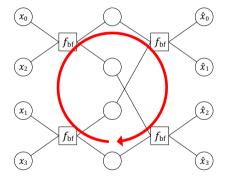
NTT with 4 coefficients

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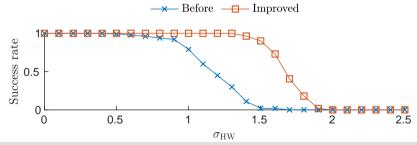
Still...



NTT with 4 coefficients
Still, shortest loops eliminated

Attack Simulations

- Leakage simulations
 - Hamming-weight with Gaussian noise
- Tripling of σ^2 (SNR)



Attacking a Real Device

Power Analysis of an ARM Cortex M4

- ASM-optimized constant-time Kyber
- Profiling: 213 univariate HW templates
- Attack: matching and run BP
- Lattice reduction for error correction
- Overall success rate: 95%



More Results

- Analyzed masking countermeasure
 - adaptation required
 - attacks still possible, but at much lower noise
- Analysis of implementation techniques
 - lazy reductions, larger input ranges
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