

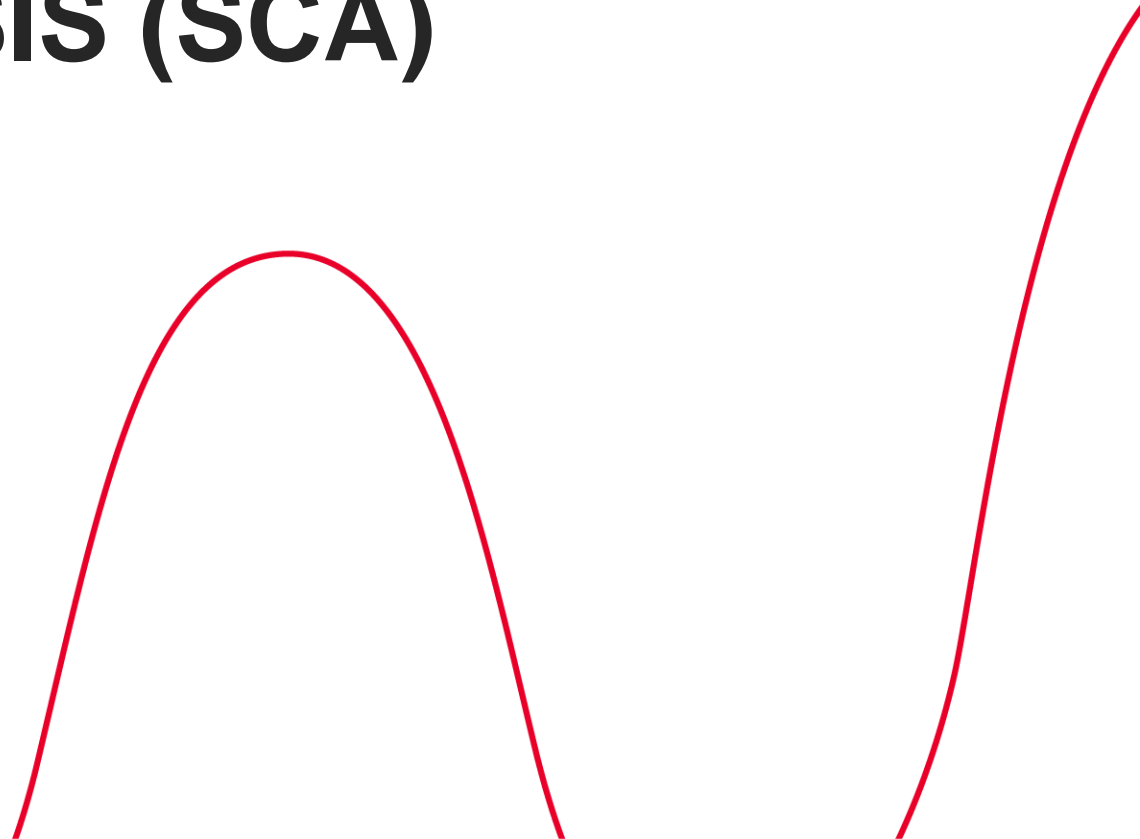
TVLA ON NTT

OPTIMIST PQC SESSION 2025
Panasayya Yalla

OUTLINE

- What is SCA?
- What is TVLA?
- What are NTT transformations?
- Relevance of NTT for PQC side channel testing?
- TVLA on NTT in ML-DSA.
- Conclusions

SIDE CHANNEL ANALYSIS (SCA)



WHAT IS A SIDE CHANNEL?

An **unintended** interface for monitoring or operating a device, resulting from its physical **implementation**.

Intended

Card reader (3x)

Keypad

Screen

USB

Printer

Speaker

GPRS / WiFi

Power / charging



Unintended

Sound

EM radiation

Power consumption

Side channel examples

Time

Power consumption

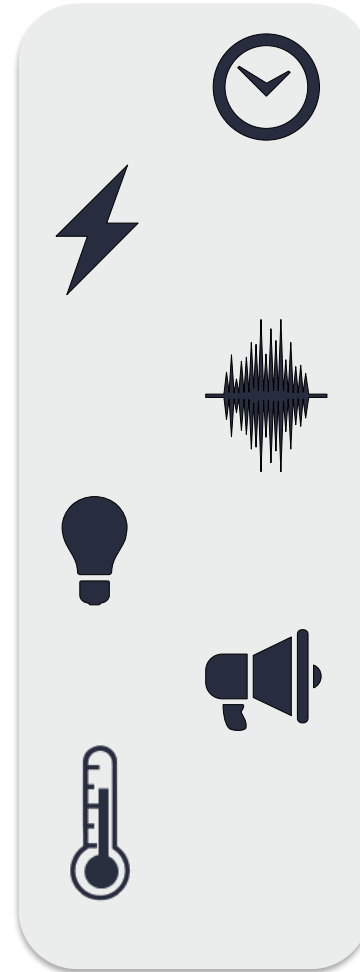
Electromagnetic radiation

Light

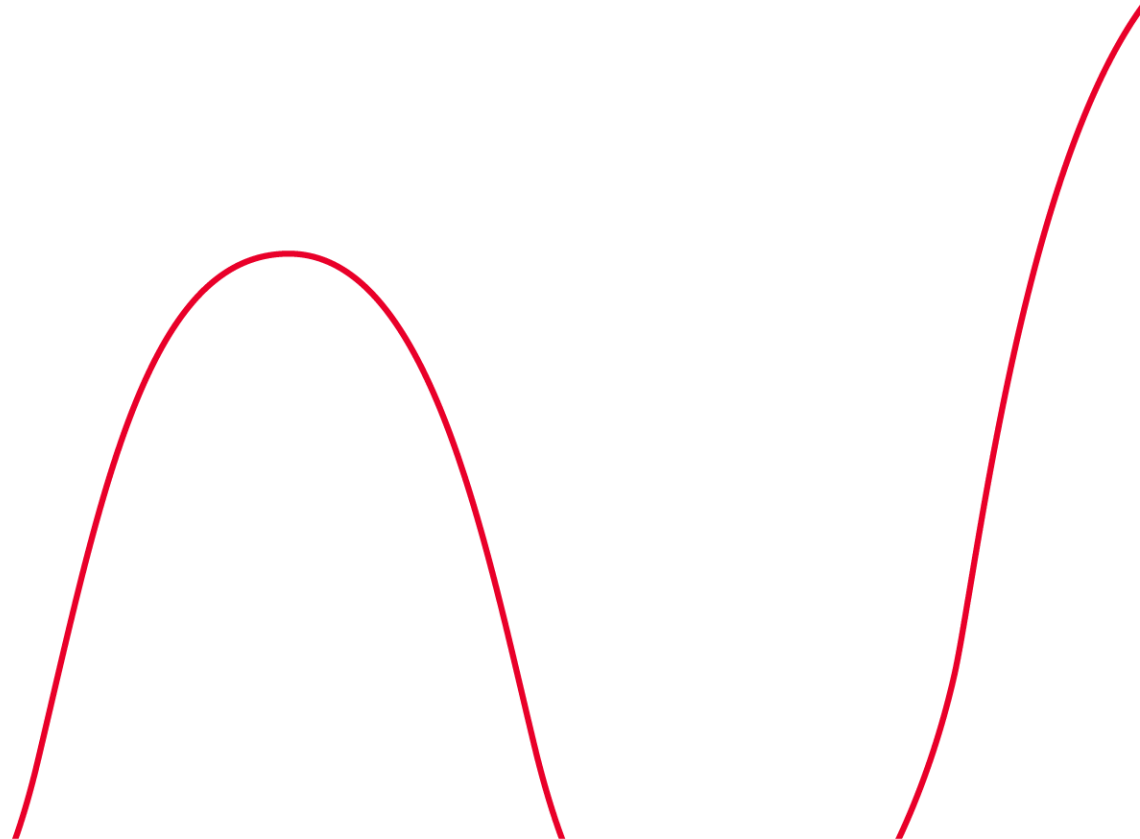
Sound

Temperature

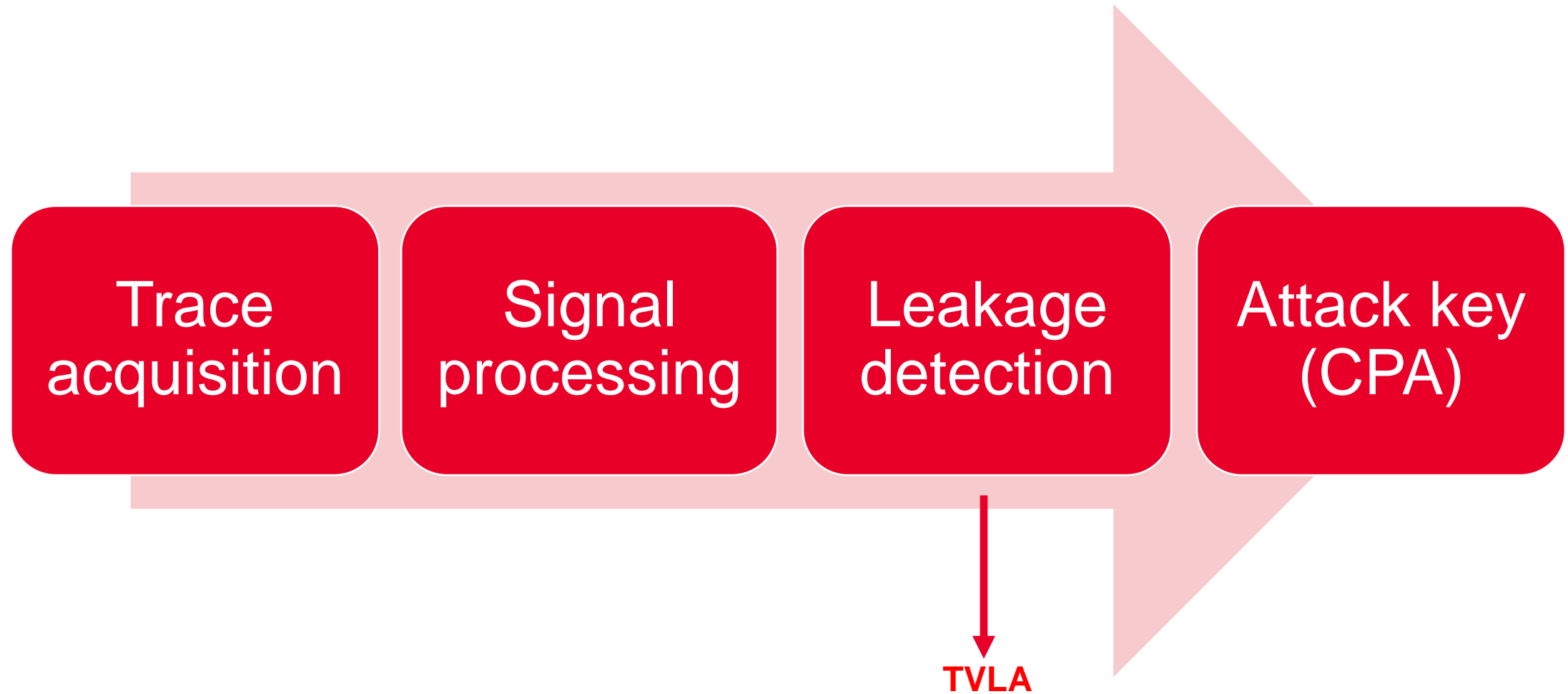
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TEST VECTOR LEAKAGE ASSESSMENT (TVLA)



GENERAL STEPS IN SCA



TVLA METHODOLOGY

Test **V**ector **L**eakage **A**ssessment is common industry practice for security evaluations in the context of SCA

Focus on information leakage, not key extraction

- Is there leakage: yes / no?
- Detecting leakage is much easier than extracting the key

Originally proposed for validation-based testing such as the FIPS 140-2 approach

Principle:

- Use an **efficient** test to detect difference in leakage
- **Maximize** the difference by crafted test vectors

TVLA highlights

- Acquire **two sets** of traces: Group A & Group B
- Make sure the statistical **distribution** of data in A & B are very **different**, e.g. random vs. constant plaintext
- Apply **Welch's T-test** to A & B:
 - Standard statistical test of equality tailored for population with **unequal variances**

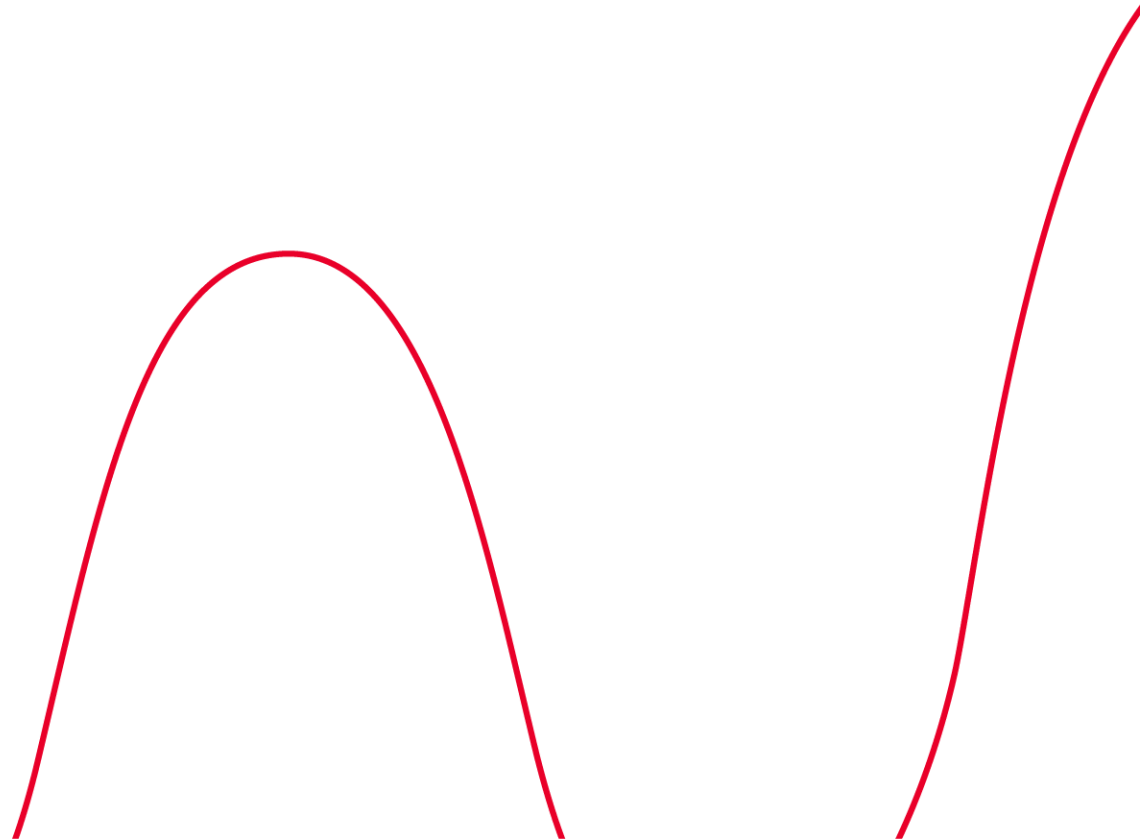
$$t(I) = \frac{X_A(I) - X_B(I)}{\sqrt{\frac{S_A^2(I)}{N_A} + \frac{S_B^2(I)}{N_B}}}$$

- T-test **spikes** → difference in signal distribution detected → **leakage** present
- Device is leaky if $t > 4.5$, for 99.999% confidence

GROUPS FOR TVLA TESTING

- Fixed vs random,
- Semi-fixed vs semi-fixed
- Biased vs random
 - Hamming Weight (HW)
 - Hamming Distance (HD)
 - Identity (ID)

NUMBER THEORETIC TRANSFORM (NTT)

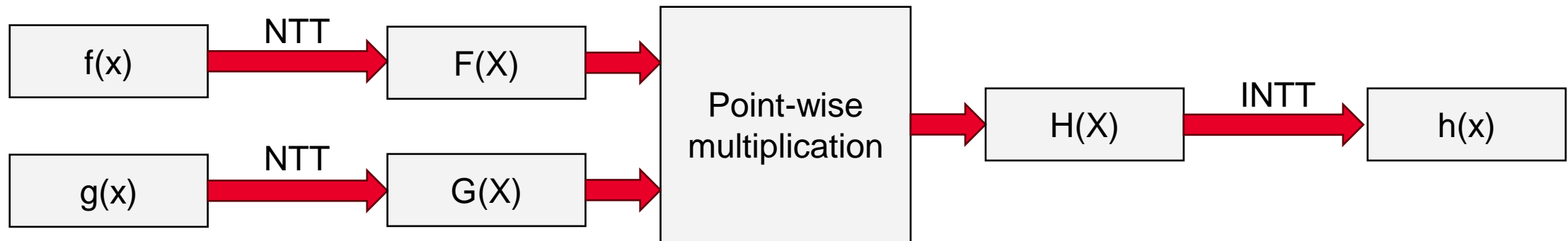


NUMBER THEORETIC TRANSFORM (NTT)

- Schoolbook multiplication $O(n^2)$

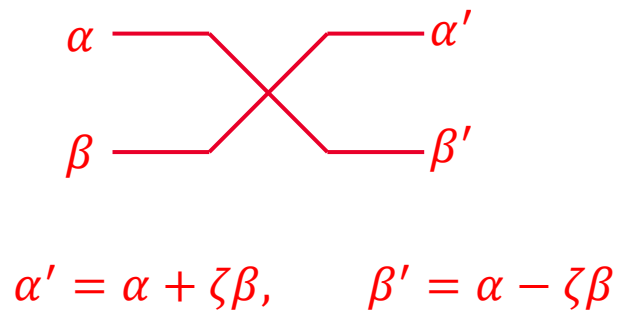
$$\begin{array}{r}
 1 + 2x + 3x^2 + 4x^3 \\
 5 + 6x + 7x^2 + 8x^3 \\
 \hline
 8x^3 + 16x^4 + 24x^5 + 32x^6 \\
 7x^2 + 14x^3 + 21x^4 + 28x^5 \\
 6x + 12x^2 + 18x^3 + 24x^4 \\
 5 + 10x + 15x^2 + 20x^3 \\
 \hline
 5 + 16x + 34x^2 + 60x^3 + 61x^4 + 52x^5 + 32x^6
 \end{array}$$

- Fast Fourier Transform (FFT) over finite fields
- Divide-and-conquer $O(n \log n)$
- Requires NTT and INTT transformations
- Workflow for computing $f(x) \cdot g(x) = h(x)$
- In NTT domain $(1X + 2X^2 + 3X^3) \cdot (4X + 5X^2 + 6X^3) = 4X + 10X^2 + 18X^3$

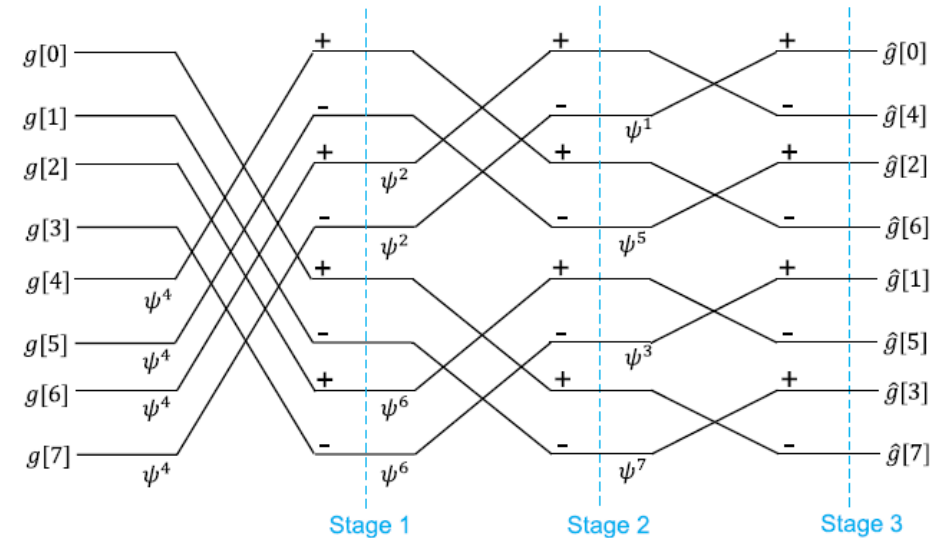


NTT BUTTERFLIES

- Divide: NTT has $\log_2 N$ stages
- Conquer: each stage has $\frac{N}{2}$ butterflies, each butterfly has 2 inputs α, β + 2 outputs α', β' + a twiddle constant ζ



Example: $N = 8$, three stages



PQC NIST STANDARD: ML-DSA

Algorithm 7 ML-DSA.Sign_internal(sk, M', rnd)

Deterministic algorithm to generate a signature for a formatted message M' .

Input: Private key $sk \in \mathbb{B}^{32+32+64+32 \cdot ((\ell+k)\text{-bitlen}(2\eta)+dk)}$, formatted message $M' \in \{0, 1\}^*$, and per message randomness or dummy variable $rnd \in \mathbb{B}^{32}$.

Output: Signature $\sigma \in \mathbb{B}^{\lambda/4 + \ell \cdot 32 \cdot (1 + \text{bitlen}(\gamma_1 - 1)) + \omega + k}$.

```

1:  $(\rho, K, tr, s_1, s_2, t_0) \leftarrow skDecode(sk)$ 
2:  $\hat{s}_1 \leftarrow NTT(s_1)$ 
3:  $\hat{s}_2 \leftarrow NTT(s_2)$ 
4:  $\hat{t}_0 \leftarrow NTT(t_0)$ 
5:  $\hat{A} \leftarrow ExpandA(\rho)$   $\triangleright A$  is generated and stored in NTT representation as  $\hat{A}$ 
6:  $\mu \leftarrow H(BytesToBits(tr) || M', 64)$   $\triangleright$  message representative that may optionally be
   computed in a different cryptographic module
7:  $\rho'' \leftarrow H(K || rnd || \mu, 64)$   $\triangleright$  compute private random seed
8:  $\kappa \leftarrow 0$   $\triangleright$  initialize counter  $\kappa$ 
9:  $(z, h) \leftarrow \perp$ 
10: while  $(z, h) = \perp$  do  $\triangleright$  rejection sampling loop
11:  $y \in R_q^\ell \leftarrow ExpandMask(\rho'', \kappa)$ 
12:  $w \leftarrow NTT^{-1}(\hat{A} \circ NTT(y))$ 
13:  $w_1 \leftarrow HighBits(w)$   $\triangleright$  signer's commitment
14:  $\triangleright HighBits$  is applied componentwise (see explanatory text in Section 7.4)
15:  $\tilde{c} \leftarrow H(\mu || w_1Encode(w_1), \lambda/4)$   $\triangleright$  commitment hash
16:  $c \in R_q \leftarrow SampleInBall(\tilde{c})$   $\triangleright$  verifier's challenge
17:  $\hat{c} \leftarrow NTT(c)$ 
18:  $\langle cs_1 \rangle \leftarrow NTT^{-1}(\hat{c} \circ \hat{s}_1)$ 
19:  $\langle cs_2 \rangle \leftarrow NTT^{-1}(\hat{c} \circ \hat{s}_2)$ 
20:  $z \leftarrow y + \langle cs_1 \rangle$   $\triangleright$  signer's response
21:  $r_0 \leftarrow LowBits(w - \langle cs_2 \rangle)$ 
22:  $\triangleright LowBits$  is applied componentwise (see explanatory text in Section 7.4)
23: if  $\|z\|_\infty \geq \gamma_1 - \beta$  or  $\|r_0\|_\infty \geq \gamma_2 - \beta$  then  $(z, h) \leftarrow \perp$   $\triangleright$  validity checks
24: else
25:  $\langle ct_0 \rangle \leftarrow NTT^{-1}(\hat{c} \circ \hat{t}_0)$ 
26:  $h \leftarrow MakeHint(-\langle ct_0 \rangle, w - \langle cs_2 \rangle + \langle ct_0 \rangle)$   $\triangleright$  Signer's hint
27:  $\triangleright MakeHint$  is applied componentwise (see explanatory text in Section 7.4)
28: if  $\|\langle ct_0 \rangle\|_\infty \geq \gamma_2$  or the number of 1's in  $h$  is greater than  $\omega$ , then  $(z, h) \leftarrow \perp$ 
29: end if
30: end if
31:  $\kappa \leftarrow \kappa + \ell$   $\triangleright$  increment counter
32: end while
33:  $\sigma \leftarrow sigEncode(\tilde{c}, z \bmod^\pm q, h)$ 
34: return  $\sigma$ 

```

Private

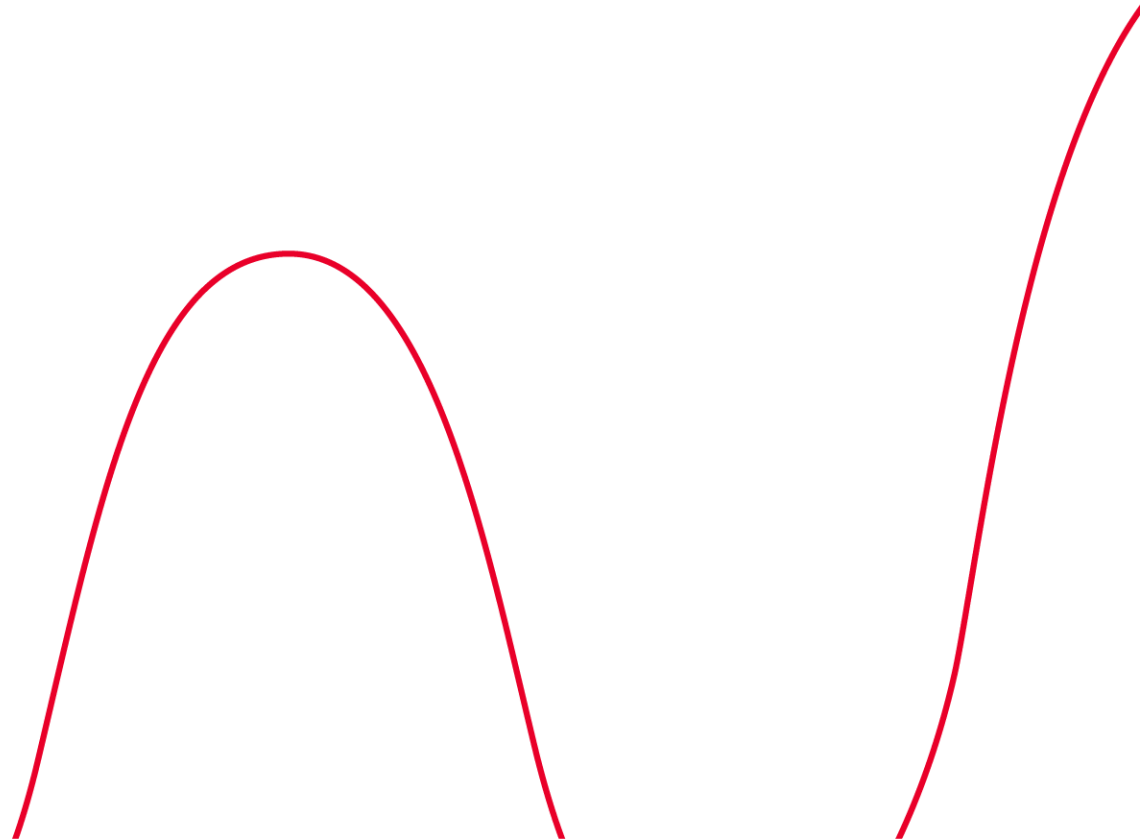
NTT and INTT

- Each polynomial has 256 coefficients
- Each coefficient is an element of mod q , where $q = 8380417 < 2^{23}$
- ML-DSA NTT has 8 stages, each stage has 128 butterflies



<https://nvlpubs.nist.gov/nistpubs/FIPS/NIST.FIPS.204.pdf>

TVLA ON NTT MODELS



TVLA FOR NTT – HW MODEL

Find NTT input such that the HW of intermediate values in i -th Stage is low (set to be h)

Steps:

1. Choose a desired HW h in stage i .
2. Find a random integer partition of h , that is, $h = h_0 + h_1 + \dots + h_{255}$ and place partitions in a vector with 256 entries $(h_0, h_1, \dots, h_{255})$.
3. As each NTT Stage is bijective, evolve the vector $(h_0, h_1, \dots, h_{255})$ from i -th stage to $(i - 1)$ -th stage, then to $(i - 2)$ -th stage, until hitting the input of NTT.

What is a good HW \square <200 is good enough

TVLA FOR NTT – HD MODEL

Find NTT input such that the HD of intermediate values before and after i -th Stage is lower than a threshold h

Do we require control the HW of XOR of values before and after a linear transformation? Yes and No

- Yes, with a “zero trick”, by setting $\beta = 0$, we have $A \begin{pmatrix} \alpha \\ 0 \end{pmatrix} = \begin{pmatrix} \alpha \\ \alpha \end{pmatrix}$, with $A = \begin{pmatrix} 1 & \zeta \\ 1 & -\zeta \end{pmatrix}$. So essentially, we gain a HD of $HD(\alpha)$ for each butterfly.
- No, bruteforce every block in a stage

Steps:

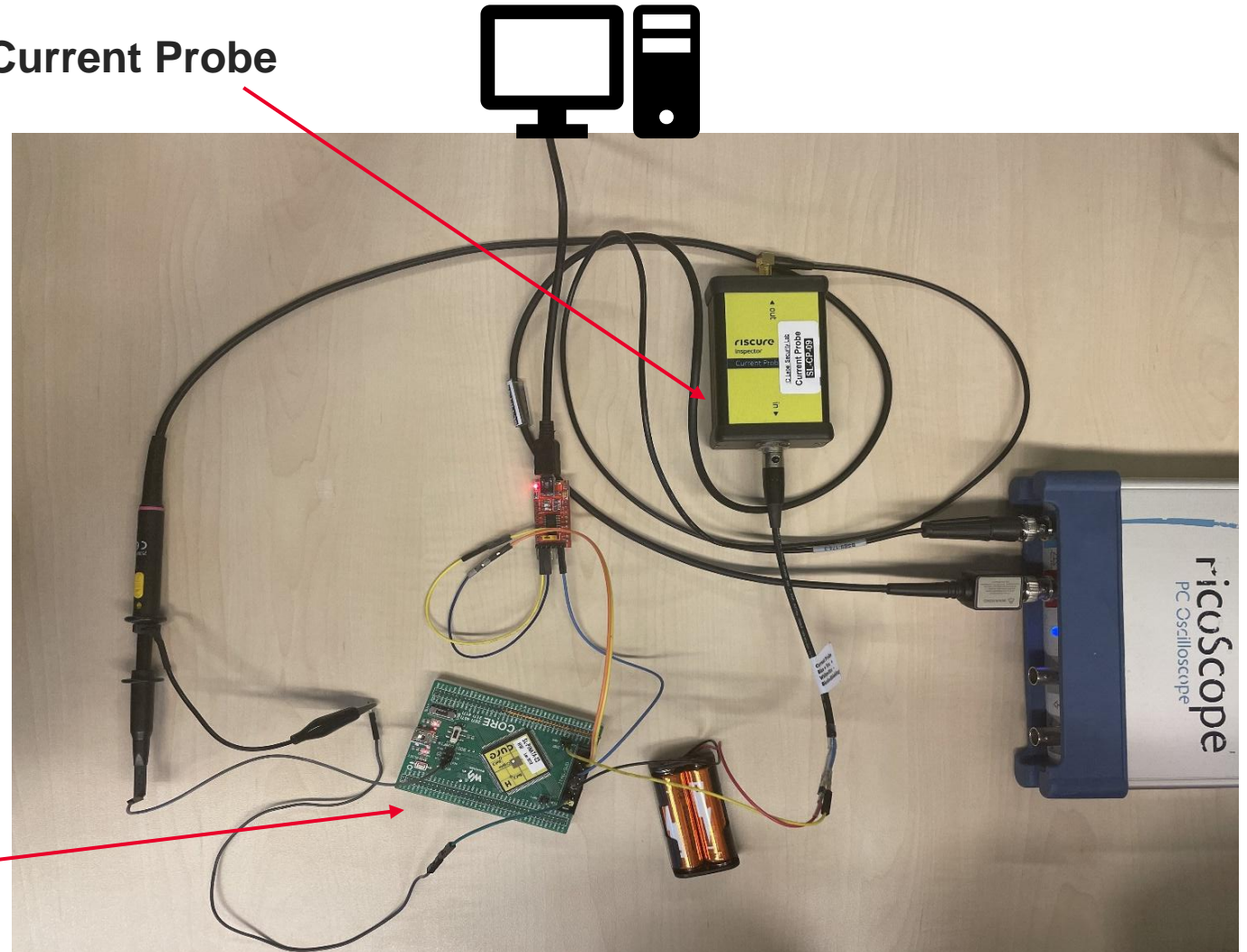
1. Fix a desired HD threshold h in stage i .
2. Find a random integer partition of h , that is, $h = h_0 + h_1 + \dots + h_{127}$ and place partitions in a vector with 256 entries, such that h_0 becomes the α for the 0-th butterfly at Stage i , that h_1 becomes the α for the 1-st butterfly at Stage $i+1$, and so on.
3. Evolve the vector back until finding the input of NTT.

EXPERIMENT SETUP

Running
standalone NTT
implementation

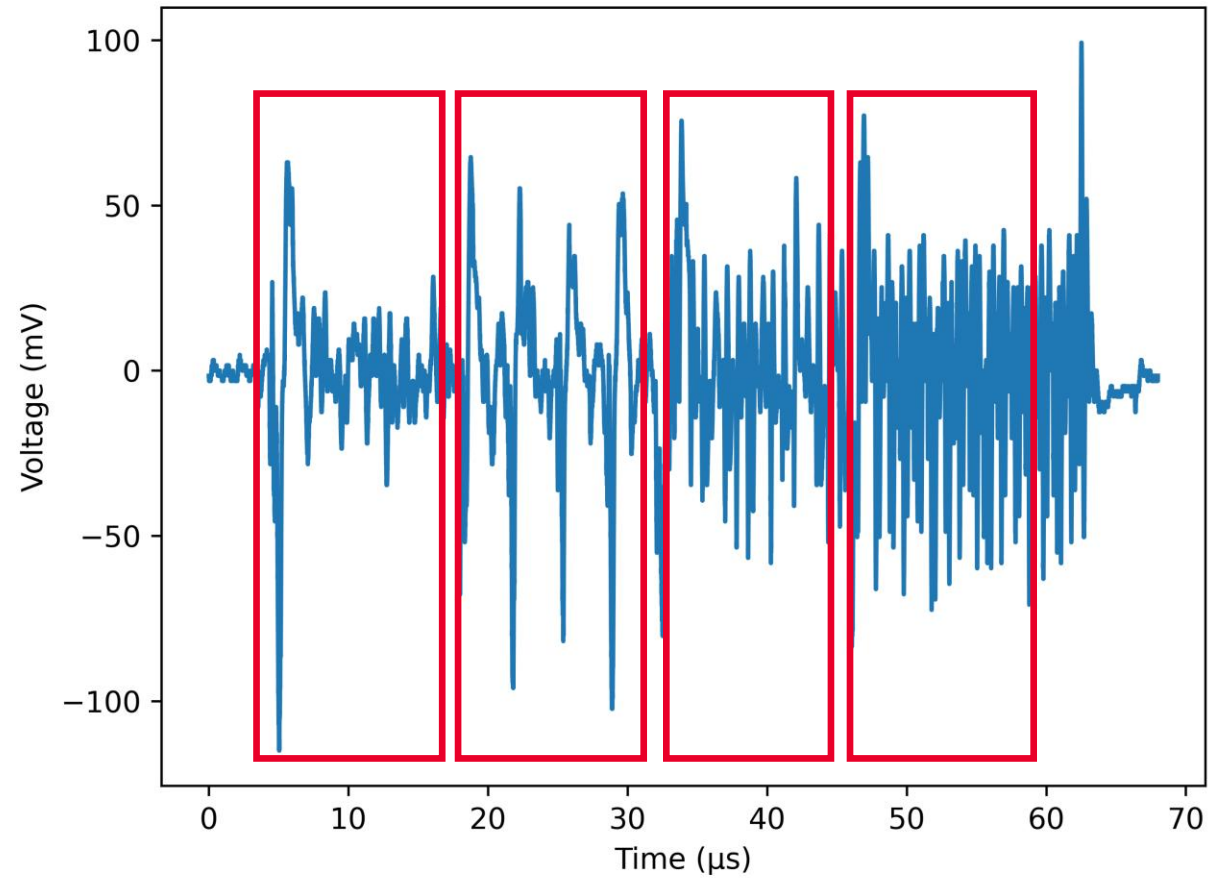
Test Target
(Piñata)

Current Probe



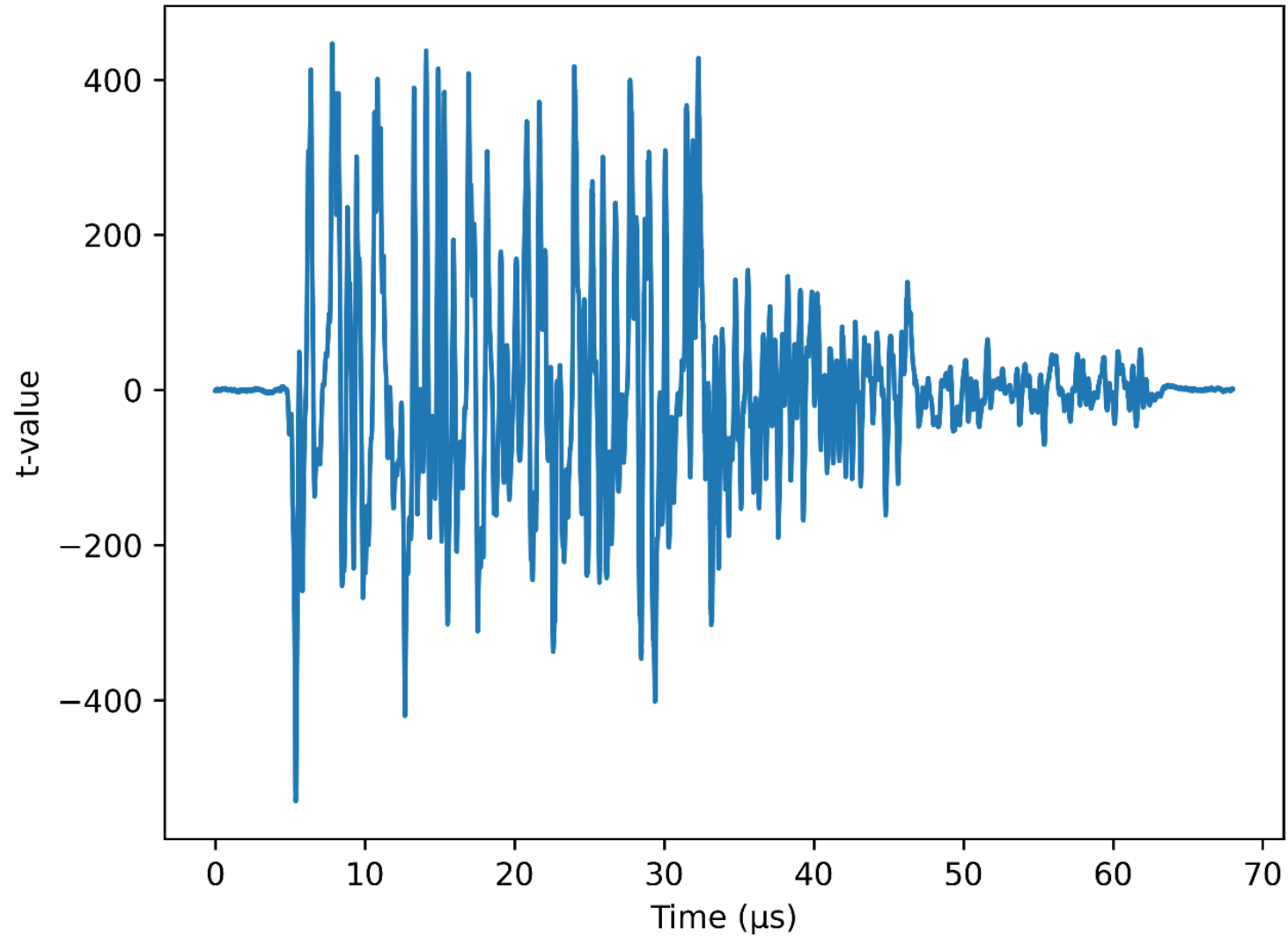
[GKS20] Greconici, D.O.C., Kannwischer, M.J., Sprenkels, A.: Compact dilithium implementations on Cortex-M3 and Cortex-M4. IACR Trans. Cryptogr. Hardw. Embed. Syst. 2021(1), 1–24 (2020)

AN EXAMPLE POWER TRACE



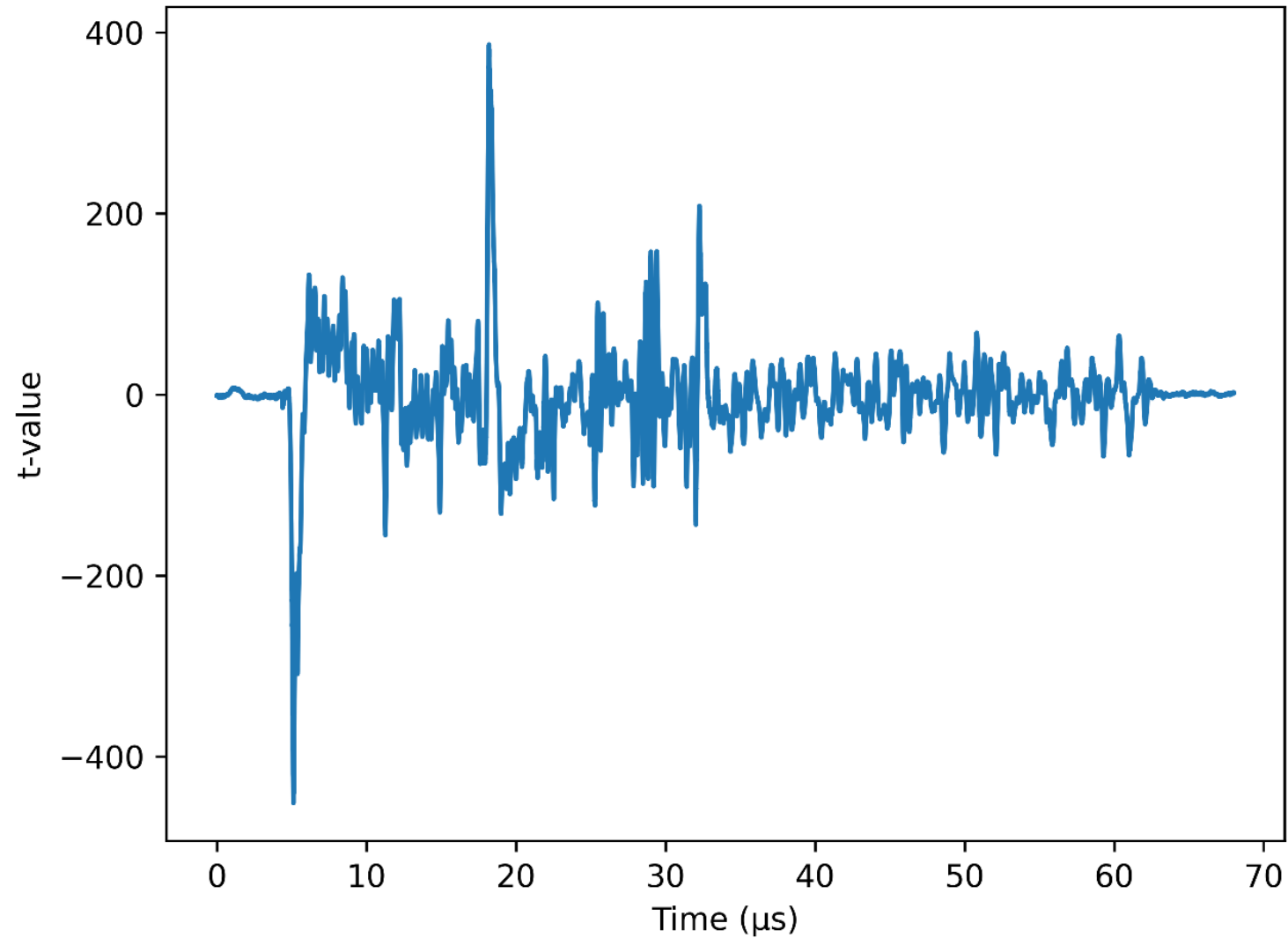
FIXED VS RANDOM TVLA (HW MODEL)

Biasing: Stage 2 with HW == 94

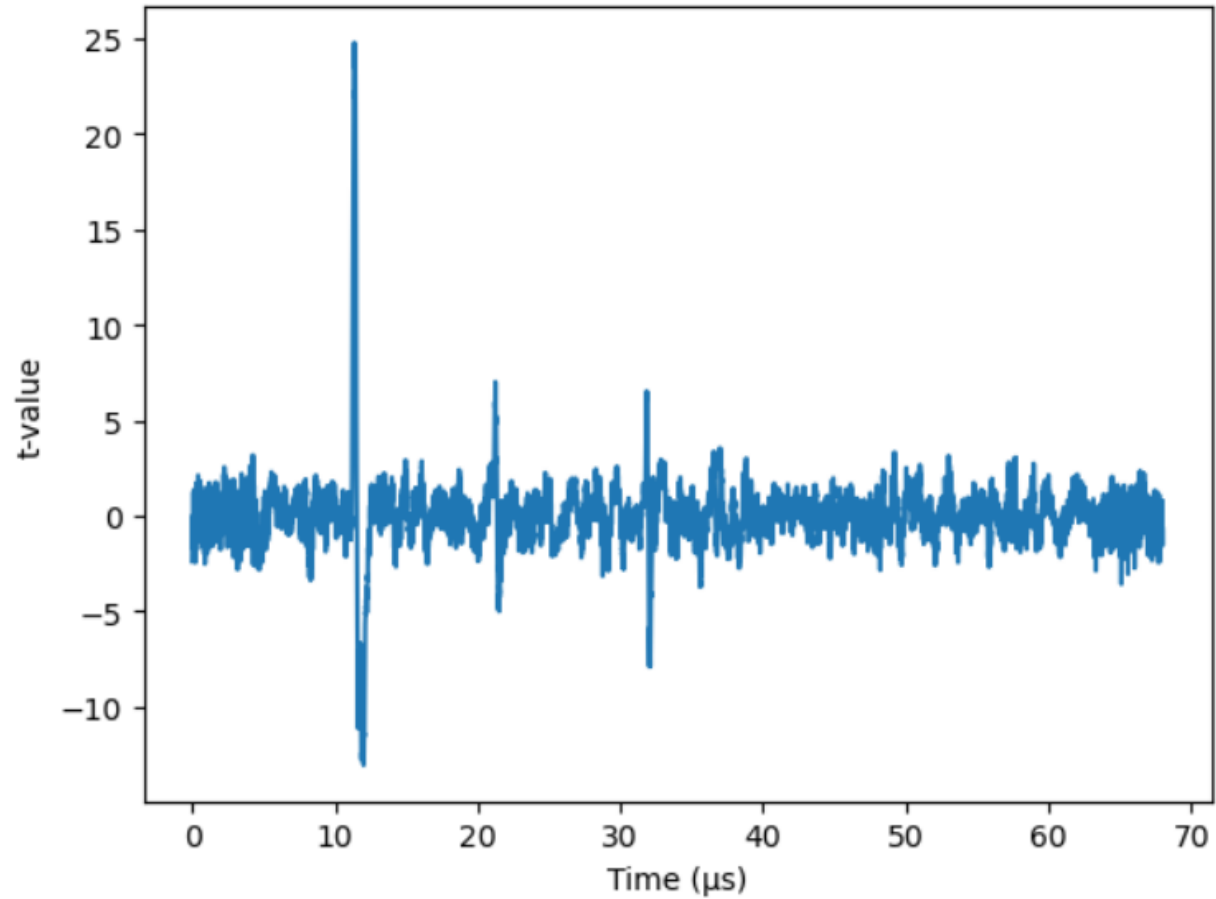


FIXED VS RANDOM TVLA (HD MODEL)

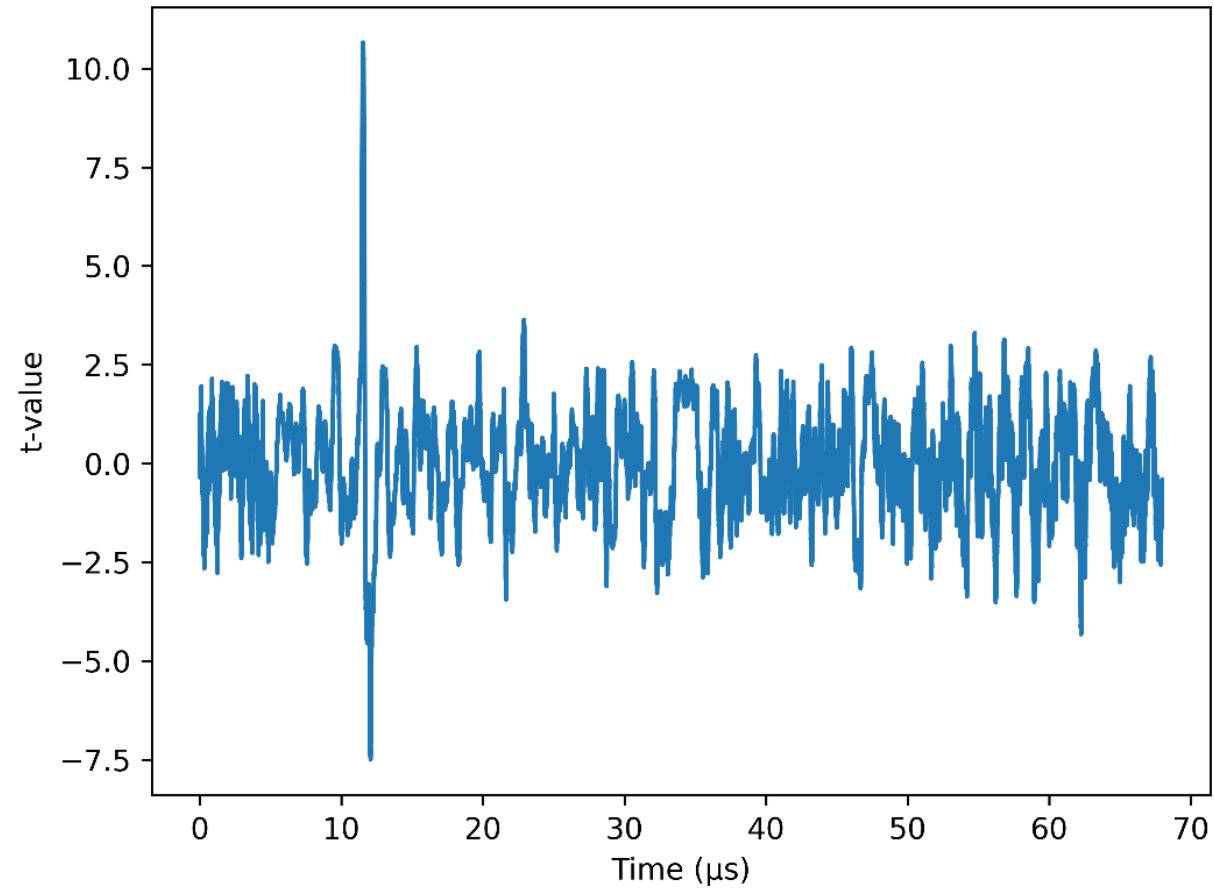
Biasing: Stage 2 with HD == 1055



SEMI-FIXED VS RANDOM TVLA (FIXING 4 BUTTERFLIES)



SEMI-FIXED VS RANDOM TVLA (FIXING 1 COEFFICIENT)



CONCLUSIONS

- NTT implementation leaks
- Leakage is mainly from memory operations
- Leakage from computing the butterflies are negligible.
- Montgomery multiplication makes it harder to generate biased vectors.

Questions?

Thank you