

# Inverse Nonlinear Fast Fourier Transform: Closing A Chapter in Quantum Signal Processing

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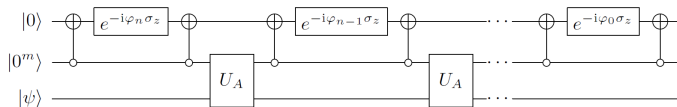
- Hongkang Ni, Rahul Sarkar, Lexing Ying, and Lin Lin. “Inverse nonlinear fast Fourier transform on  $SU(2)$  with applications to quantum signal processing.” arXiv preprint arXiv:2505.12615 (2025).
- <https://github.com/qsppack/QSPPACK>

- 1 Introduction
- 2 QSP, GQSP, and NLFT
- 3 Layer Stripping: Stable or Not?
- 4 INLFFT: The Fast Algorithm
- 5 Conclusion

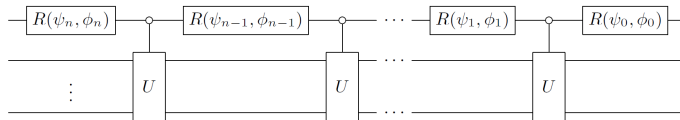
# Matrix Function Computation: QSP<sup>1</sup> and GQSP<sup>2</sup>

Numerous applications:

- Hamiltonian simulation
- Linear system solver
- Eigenvalue problems
- Gibbs states preparation
- ...



QSP circuit



GQSP circuit

<sup>1</sup>(Low, Chuang, PRL 2017, QIP'17)

<sup>2</sup>(Motlagh, Wiebe, PRX Quantum 2024)

# QSP Phase Factor Finding Problem

- Qubitization technique: only need to find phase factors for  $2 \times 2$  matrix!
- Single qubit Pauli matrices

$$\sigma_X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \quad \sigma_Z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

- For  $x = \cos \theta \in [-1, 1]$ , rotation matrix

$$W(x) = e^{i\theta\sigma_X} = \begin{pmatrix} x & i\sqrt{1-x^2} \\ i\sqrt{1-x^2} & x \end{pmatrix}$$

## Problem (QSP phase factor finding)

Given an even (or odd) real target polynomial  $f(x)$  with  $\|f\|_{L^\infty[-1,1]} \leq 1 - \eta$ , how to find a sequence  $\Phi = (\varphi_0, \dots, \varphi_n)$  such that  $f(x) = \text{Im}[P(x)]$ , where  $P(x)$  is defined by

$$\begin{pmatrix} P(x) & * \\ * & * \end{pmatrix} = e^{i\varphi_0\sigma_Z} W(x) e^{i\varphi_1\sigma_Z} W(x) \dots e^{i\varphi_{n-1}\sigma_Z} W(x) e^{i\varphi_n\sigma_Z}?$$

- Two-parameter single-qubit rotations

$$R(\psi, \phi) := \begin{pmatrix} \cos \psi & e^{i\phi} \sin \psi \\ -e^{-i\phi} \sin \psi & \cos \psi \end{pmatrix} \otimes I$$

- Lift the parity constraint on target polynomial

## Problem (GQSP phase factor finding)

Given a target polynomial  $b(\cdot)$  of degree  $n$  such that  $\|b\|_{L^\infty(\mathbb{T})} \leq 1$ , how to find a sequence of angles

$$\{(\psi_k, \phi_k)\}_{k=0}^n$$

such that

$$\begin{pmatrix} * & b(U) \\ * & * \end{pmatrix} = R(\psi_0, \phi_0) \begin{pmatrix} U & \\ & I \end{pmatrix} \cdots \begin{pmatrix} U & \\ & I \end{pmatrix} R(\psi_n, \phi_n)?$$

In the early stages of QSP development..

# Toward the first quantum simulation with quantum speedup

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## Section H.3:

...However, this computation is difficult in practice, so we can only carry it out for very small instances. Specifically, we found the time required to calculate the angles to be **prohibitive for values of  $M$  greater than about 32**...It is a natural open problem to give a more practical method for computing the angles

## Key problem 1: Efficiency

- Iterative methods:  $\tilde{\mathcal{O}}(n) \sim \tilde{\mathcal{O}}(n^2)$ . Fails in fully-coherent regime (small  $\eta$ ).
- Direct methods:

Algorithm	Time complexity	Space complexity
Riemann-Hilbert <sup>1</sup>	$\tilde{\mathcal{O}}(n^4 + n\eta^{-1} \log^2(\epsilon^{-1}))$	$\mathcal{O}(n^2)$
Half-Cholesky <sup>2</sup>	$\tilde{\mathcal{O}}(n^2 + n\eta^{-1} \log^2(\epsilon^{-1}))$	$\mathcal{O}(n)$
Layer stripping <sup>3,4</sup>	$\tilde{\mathcal{O}}(n^2 + n\eta^{-1} \log^2(\epsilon^{-1}))$	$\mathcal{O}(n)$
Inverse nonlinear FFT	$\tilde{\mathcal{O}}(n + n\eta^{-1} \log^2(\epsilon^{-1}))$	$\mathcal{O}(n)$

- Easily compute for  $n \sim 10^6$

<sup>1</sup>(Alexis, Lin, Mnatsakanyan, Thiele, Wang, CPAM 2026, QIP'25)

<sup>2</sup>(Ni, Ying, arXiv:2410.06409)

<sup>3</sup>(Y.-J. Tsai, Thesis, 2005)

<sup>4</sup>(Gilyén, Su, Low, Wiebe, STOC 2019, QIP'19)



## Key problem 2: Numerical Stability

What is numerical stability?

- **Intuition:** Small errors should *not* grow uncontrollably as the algorithm proceeds.
- **Practical perspective:** Truncation errors negligible on common devices  
e.g. 52 bits for IEEE double precision
- **Theoretical perspective:** Bit requirement grows only *poly-logarithmically* in problem parameters.
- **In our setting:** Bit requirement scales as

$$r = \text{polylog}(n, \eta^{-1}, \epsilon^{-1}),$$

## Key problem 2: Numerical Stability

Iterative methods:

- Stable when the algorithm converges.

Direct methods:

- Step 1: Complementary polynomial finding:
  - Root finding approach<sup>1</sup>: Unstable
  - Weiss algorithm<sup>2</sup>: Stable
- Step2: Layer stripping: ?

### Problem

- *Is it possible to prove the layer stripping part of the direct methods is stable?*  
*Yes! (Under certain conditions)*
- *Is it possible to develop a more efficient layer stripping type method that preserves stability?*  
*Yes!*

<sup>1</sup>(Gilyén, Su, Low, Wiebe, STOC 2019, QIP'19), (Haah, Quantum 2019)

<sup>2</sup>(Alexis, Lin, Mnatsakanyan, Thiele, Wang, CPAM 2026, QIP'25)

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# Nonlinear Fourier Transform (NLFT)

- Given a compactly supported sequence  $\gamma = (\gamma_k)_{0 \leq k \leq n}$  of complex numbers, we define its Nonlinear Fourier Transform<sup>1</sup> (NLFT) as

$$\widehat{\gamma} := \begin{pmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{pmatrix} = \prod_{k=0}^n \left[ \frac{1}{\sqrt{1 + |\gamma_k|^2}} \begin{pmatrix} 1 & \gamma_k z^k \\ -\overline{\gamma_k} z^{-k} & 1 \end{pmatrix} \right]$$

Here,  $a^*(z) := \overline{a(\overline{z^{-1}})}$ .

- The image of NLFT is

$$\mathcal{S} = \{(a, b) : aa^* + bb^* = 1, 0 < a(\infty) < \infty\}.$$

## Problem (Inverse NLFT)

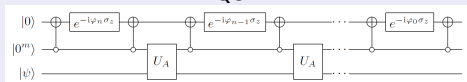
Given a Laurent polynomial  $b(z)$  satisfying  $\|b\|_{L^\infty(\mathbb{T})} \leq 1$ , determine a compactly supported sequence  $\gamma$  such that

$$\widehat{\gamma} = \begin{pmatrix} * & b(z) \\ * & * \end{pmatrix}.$$

<sup>1</sup>(Tao, Thiele, arXiv:1201.5129), (Y.-J. Tsai, Thesis, 2005), (Alexis, Mnatsakanyan, Thiele, 2024)

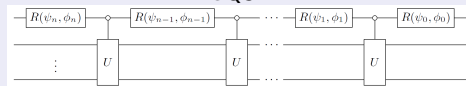
# Converting (G)QSP to NLFT<sup>1</sup>

QSP



$$e^{i\varphi_0\sigma_z} W(x) e^{i\varphi_1\sigma_z} W(x) \dots W(x) e^{i\varphi_n\sigma_z}.$$

GQSP



$$R(\psi_0, \phi_0) \cdot [U] \cdot \dots \cdot [U] \cdot R(\psi_n, \phi_n)$$

Nonlinear Fourier Transform

$$\begin{pmatrix} * & b(z) \\ * & * \end{pmatrix} = \prod_{k=0}^n \left[ \frac{1}{\sqrt{1 + |\gamma_k|^2}} \begin{pmatrix} 1 & \gamma_k z^k \\ -\overline{\gamma_k} z^{-k} & 1 \end{pmatrix} \right]$$

<sup>1</sup>See details in Section 3 of our paper. See also (Laneve, arXiv:2503.03026)

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# The Layer Stripping Algorithm

## QSP<sup>1,2</sup>

- Given  $\text{Im}[P(x)]$
- Find complementary polynomials  $\text{Re}[P(x)]$  and  $Q(x)$  such that

$$\begin{bmatrix} P(x) & iQ(x)\sqrt{1-x^2} \\ iQ^*(x)\sqrt{1-x^2} & P^*(x) \end{bmatrix}$$

is unitary for  $x \in [-1, 1]$ .

- The first phase factor  $\phi_0$  can be obtained from the leading coefficient of  $P$  and  $Q$
- Undo the first rotation and retrieve the following  $\phi_k$ 's sequentially using the same method

## NLFT<sup>3,4</sup>

- Given  $b(z)$
- Find complementary Laurent polynomial  $a(z)$  such that

$$\begin{bmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{bmatrix}$$

is unitary for  $z \in \mathbb{T}$ .

- The first coefficient  $\gamma_0$  can be obtained from the leading coefficient of  $a$  and  $b$
- Undo the first rotation and retrieve the following  $\gamma_k$ 's sequentially using the same method

<sup>1</sup>(Gilyén, Su, Low, Wiebe, STOC 2019, QIP'19)

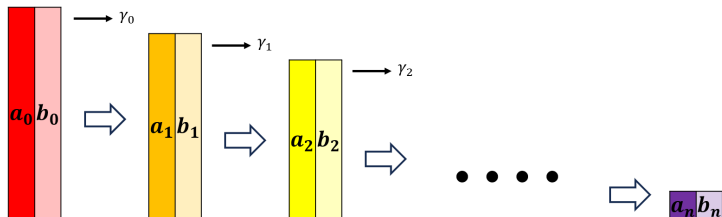
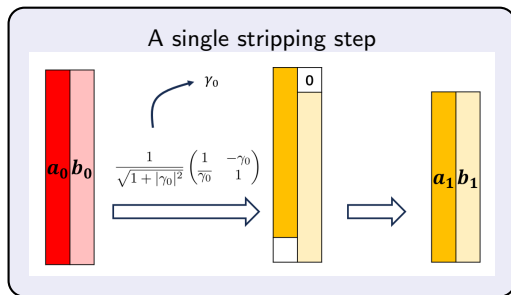
<sup>2</sup>(Haah, Quantum 2019)

<sup>3</sup>(Tao, Thiele, arXiv:1201.5129)

<sup>4</sup>(Alexis, Mnatsakanyan, Thiele, RMC 2024)

# Layer Stripping (Matrix Form)

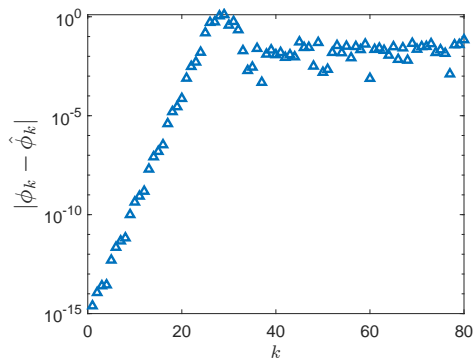
- $\mathbf{a}_0, \mathbf{b}_0$  are the coefficient vectors of  $a(z), b(z)$





# Instability of layer stripping

- The numerical error may accumulate exponentially during this sequential algorithm.
- Layer stripping works under  $\tilde{\mathcal{O}}(n)$  bits precision floating point arithmetic.<sup>1</sup>
- Instability justified by the following randomly generated example:



Error accumulation of the phase factors for an 80-degree polynomial pair  $(a(z), b(z))$

<sup>1</sup>(Haah, Quantum 2019)

# Stable Layer Stripping Algorithm

- Matrix completion problem: Given  $b(z)$ , find  $a(z)$  to complete  $\begin{pmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{pmatrix}$
- Combinatorially many choices<sup>1</sup> of  $a(z)$
- Key insight<sup>2</sup>: let  $a^*(z) := \overline{a(\overline{z^{-1}})}$  be an **outer** function: roughly,  $a^*$  has no zeros in the closed unit disk
- This outer  $a^*$  can be constructed efficiently and stably using the Weiss algorithm

## Theorem

When  **$a^*(z)$  is outer**, the layer stripping algorithm for the inverse NLFT problem is numerically stable.

<sup>1</sup>(Gilyén, Su, Low, Wiebe, STOC 2019, QIP'19), (Haah, Quantum 2019), (Wang, Dong, Lin, Quantum 2022)

<sup>2</sup>(Alexis, Lin, Mnatsakanyan, Thiele, Wang, CPAM 2026, QIP'25)

- Use the displacement structured matrix theory<sup>1</sup>
- Consider the unique matrix  $K$  satisfying

$$K - ZKZ^\dagger = \mathbf{a}_0\mathbf{a}_0^\dagger + \mathbf{b}_0\mathbf{b}_0^\dagger, \quad Z = \begin{pmatrix} 0 & & & & \\ 1 & 0 & & & \\ & \ddots & \ddots & & \\ & & \ddots & 1 & 0 \end{pmatrix}$$

- $K$  has the triangular factorization

$$K = UU^\dagger, \quad U = \begin{array}{|c|} \hline \text{Red bar } a_0 \\ \text{Orange bar } a_1 \\ \text{Yellow bar } a_2 \\ \text{Light Green bar } \\ \text{Green bar } \\ \text{Blue bar } \\ \text{Dark Blue bar } \\ \text{Purple bar } a_n \\ \hline \end{array}$$

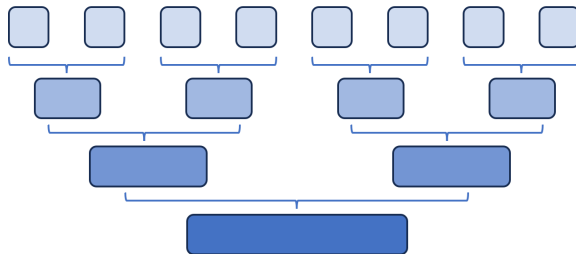
- Doing layer stripping  $\approx$  Factorizing matrix  $K$  (using Schur's algorithm)
- $K$  is well-conditioned when  $a^*(z)$  is outer.

<sup>1</sup>(Kailath, Sayed, SIAM Review 1995)

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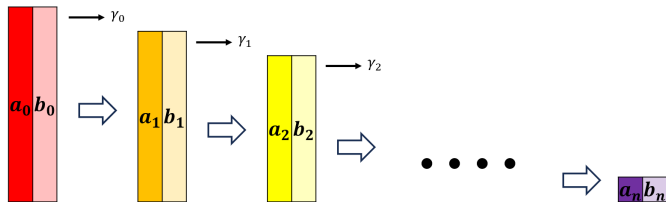
$$\begin{pmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{pmatrix} = \prod_{k=0}^n \left[ \frac{1}{\sqrt{1 + |\gamma_k|^2}} \begin{pmatrix} 1 & \gamma_k z^k \\ -\overline{\gamma_k} z^{-k} & 1 \end{pmatrix} \right]$$

- Q: How to calculate the polynomial matrix product more efficiently?
- A: Divide and conquer + FFT!

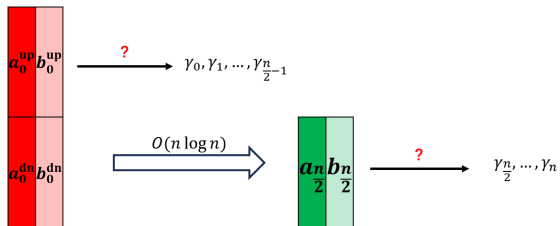


# INLFFT

- Layer stripping:  $\mathcal{O}(n^2)$



- INLFFT:  $\mathcal{O}(n \log^2 n)$



$$\begin{pmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{pmatrix} = \prod_{k=0}^n \left[ \frac{1}{\sqrt{1+|\gamma_k|^2}} \begin{pmatrix} 1 & \gamma_k z^k \\ -\bar{\gamma}_k z^{-k} & 1 \end{pmatrix} \right]$$

Split the product and take inverse ( $m = \lceil \frac{n}{2} \rceil$ ):

$$\prod_{k=m-1}^0 \left[ \frac{1}{\sqrt{1+|\gamma_k|^2}} \begin{pmatrix} 1 & -\gamma_k z^k \\ \bar{\gamma}_k z^{-k} & 1 \end{pmatrix} \right] \begin{pmatrix} a(z) & b(z) \\ -b^*(z) & a^*(z) \end{pmatrix} = \begin{pmatrix} a_m(z) & z^m b_m(z) \\ -z^{-m} b_m^*(z) & a_m^*(z) \end{pmatrix}$$

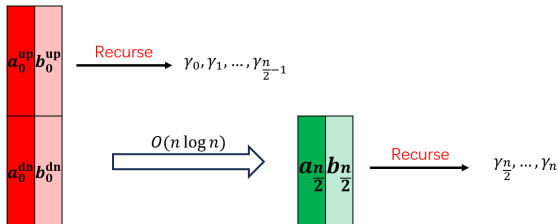
### Algorithm Sketch

- ① Calculate the first half of the NLFT coefficients  $\gamma_0, \dots, \gamma_{m-1}$ .
- ② Use the divide-and-conquer type fast algorithm to calculate the product  $\prod_{k=m-1}^0 \left[ \frac{1}{\sqrt{1+|\gamma_k|^2}} \begin{pmatrix} 1 & -\gamma_k z^k \\ \bar{\gamma}_k z^{-k} & 1 \end{pmatrix} \right]$ .
- ③ Construct  $(a_m^*, b_m)$  by the above formula.
- ④ Retrieve the last half of the NLFT coefficients  $\gamma_m, \dots, \gamma_n$  from  $(a_m^*, b_m)$ .

## Algorithm Sketch

- ① Calculate the first half of the NLFT coefficients  $\gamma_0, \dots, \gamma_{m-1}$ .
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- ③ Construct  $(a_m^*, b_m)$  by the above formula.
- ④ Retrieve the last half of the NLFT coefficients  $\gamma_m, \dots, \gamma_n$  from  $(a_m^*, b_m)$ .

- Key insight:  $\gamma_0, \dots, \gamma_{m-1}$  only depends on the first half of  $\mathbf{a}_0$  and  $\mathbf{b}_0$
- Step 1 and 4 can be done recursively





## Theorem

When  $a^*(z)$  is *outer*, the INLFFT algorithm is numerically stable.

See Section 5.5 in our paper for proof.

- 1 Introduction
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- **Algorithmic Contribution:**

- Developed INLFFT: A fast, divide-and-conquer implementation of the Inverse NLFT.
- Achieved near optimal complexity of  $\tilde{O}(n)$ , improving upon the previous  $\tilde{O}(n^2)$  barrier.

- **Theoretical Contribution:**

- We resolved the long-standing open question regarding the numerical stability of layer stripping.
- We further established the stability result of INLFFT.

- **Closing a Chapter in (G)QSP Phase Factoring Finding**

- We unify QSP and GQSP within the NLFT framework
- This leads to phase factor finding algorithms that are both efficient and stable.

- **Further Directions:**

- Multi-variable QSP?
- NLFT in higher dimensions?

Thank you for your attention!