

Encoding **Discrete Fourier Transform** as **Quantized Tensor Train**

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Joint work with Miles Stoudenmire, Steve White, Michael Lindsey

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Background

Quantized tensor train (QTT) is a type of tensor network with linear geometry, which represents an underlying function in a multi-scale fashion.

It was found useful in many fields, including but not limited to:

- Representing bosonic systems [1]
- Image compression [2]
- Differential equations, e.g. Navier-Stokes [3], Vlasov-Poisson [4]
- Encoding orbitals for computational chemistry [5]

Also closely related to “entanglement structure” of quantum algorithms!

[1] Eric Jeckelmann, Steven R. White, *Density-matrix renormalization-group study of the polaron problem in the Holstein model*, Phys. Rev. B 57, 6376–6385 (1998)

[2] Jose I. Latorre, *Image compression and entanglement*, quant-ph/0510031 (2005)

[3] Nikita Gourianov, et.al., *A quantum-inspired approach to exploit turbulence structures*, Nature Computational Science 2, 30–37 (2022)

[4] Erika Ye, Nuno F. G. Loureiro, *Quantum-inspired method for solving the Vlasov-Poisson equations*, Phys. Rev. E 106, 035208 (2022)

[5] Nicolas Jolly, Yurriel Nunez Fernandez, Xavier Waintal, *Tensorized orbitals for computational chemistry*, 2308.03508 (2023)

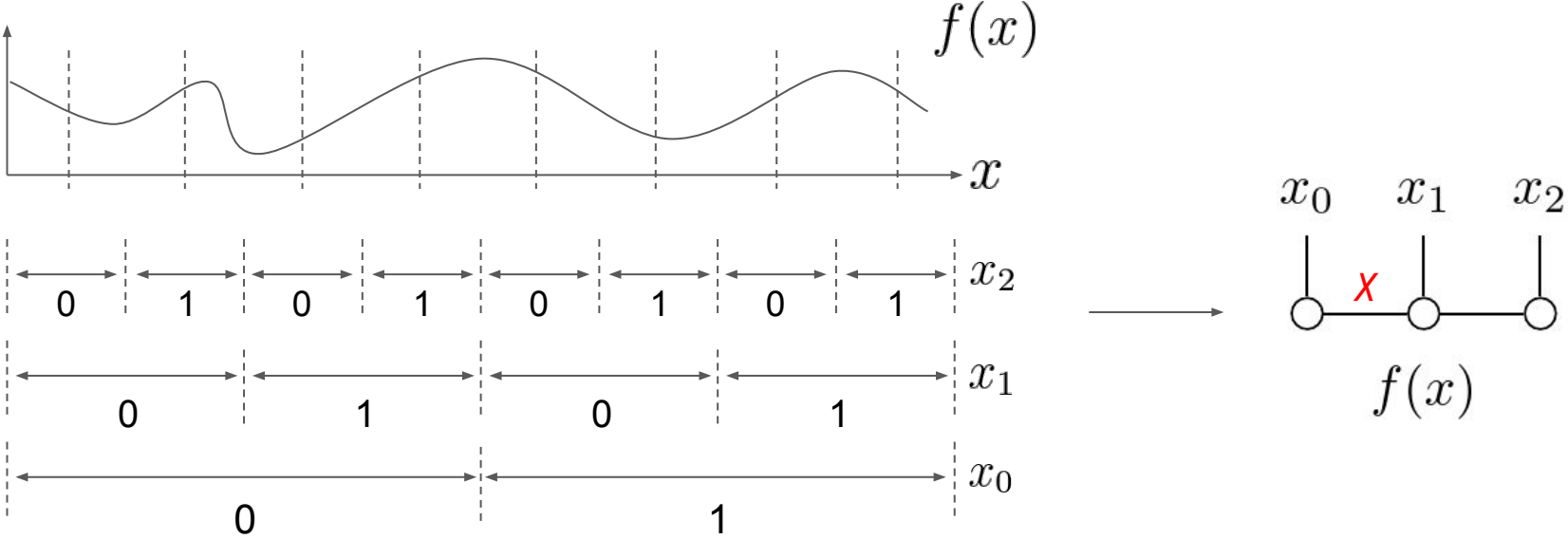
Background

Recently it was found that **discrete Fourier transform (DFT)** can be encoded as an efficient QTT [1].

This could be useful for many QTT applications since DFT is a fundamental algorithm primitive.

This talk: recent results on constructing DFT in QTT format.

Quantized Tensor Train: 1D functions



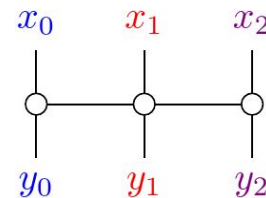
Efficient criteria: bond-dimension $\chi = \text{poly}(n)$ or even $\chi = O(1)$

Space savings: $2^n \rightarrow O(n\chi^2)$

Quantized Tensor Train: operators

	y_2	y_1	y_0					
x_2	a_{11}	a_{12}	a_{13}	a_{14}	a_{15}	a_{16}	a_{17}	a_{18}
x_1	a_{21}	a_{22}	a_{23}	a_{24}	a_{25}	a_{26}	a_{27}	a_{28}
x_0	a_{31}	a_{32}	a_{33}	a_{34}	a_{35}	a_{36}	a_{37}	a_{38}
	a_{41}	a_{42}	a_{43}	a_{44}	a_{45}	a_{46}	a_{47}	a_{48}
	a_{51}	a_{52}	a_{53}	a_{54}	a_{55}	a_{56}	a_{57}	a_{58}
	a_{61}	a_{62}	a_{63}	a_{64}	a_{65}	a_{66}	a_{67}	a_{68}
	a_{71}	a_{72}	a_{73}	a_{74}	a_{75}	a_{76}	a_{77}	a_{78}
	a_{81}	a_{82}	a_{83}	a_{84}	a_{85}	a_{86}	a_{87}	a_{88}

operator \rightarrow



Space savings: $2^{2n} \rightarrow O(n\chi^2)$

Similar scheme for multilinear map & high-dimensional functions

Discrete Fourier Transform

Defined as the following matrix, where $\omega = e^{-\frac{2\pi i}{N}}$ and $N = 2^n$

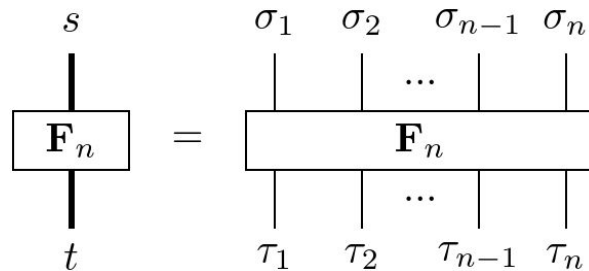
$$\mathbf{F}_n(s, t) = e^{-\frac{2\pi i s t}{N}} = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \omega^3 & \dots & \omega^{N-1} \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots & \omega^{2(N-1)} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots & \omega^{3(N-1)} \\ \vdots & \vdots & \vdots & \vdots & & \vdots \\ 1 & \omega^{N-1} & \omega^{2(N-1)} & \omega^{3(N-1)} & \dots & \omega^{(N-1)(N-1)} \end{bmatrix}$$

Discrete Fourier Transform

Can partition variables into binary strings:

$$\mathbf{F}_n(s, t) = e^{-\frac{2\pi i s t}{N}}$$

$$s = \sum_{k=1}^n 2^{n-k} \sigma_k, \quad t = \sum_{k=1}^n 2^{k-1} \tau_k$$

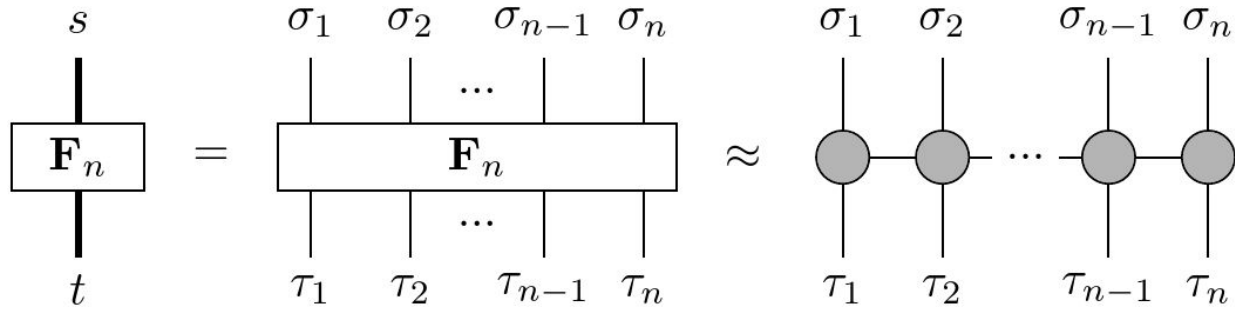


Notice: σ_1 is the most significant bit, but τ_1 is the least significant bit.

Why? Only this ordering provides us a low-rank QTT of DFT.
Normal ordering will give maximal QTT rank.

DFT in QTT

Main result: DFT can be approximated as low-rank QTT with small error



Proof by:

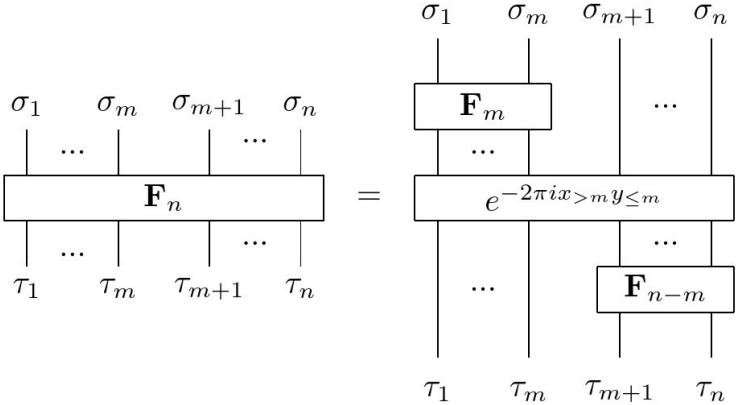
1. Schmidt decomposition (Frobenius norm error bound) [1]
2. Interpolative decomposition (Frobenius norm + entry-wise error bound + analytical construction) [2]

[1] JC, E.M. Stoudenmire, and Steven R. White, *Quantum Fourier transform has small entanglement*, PRX Quantum (2023) 4, 040318


[2] JC, Michael Lindsey, *Direct interpolative construction of the discrete Fourier transform as a matrix product operator*, arXiv:2404.03182 (2024)

Step 1: Recursive decomposition of DFT

DFT = products of smaller DFTs with some phase modifications
 Closely related to the Cooley–Tukey algorithm (fast Fourier transform)



$$\begin{aligned}
 & \mathbf{F}_n(s, t) \\
 &= e^{-\pi i \sum_{k,l=1}^n 2^{-k} 2^l \sigma_k \tau_l} \\
 &= e^{-\pi i \sum_{k,l=1}^m 2^{-k} 2^l \sigma_k \tau_l} \times e^{-\pi i \sum_{k=m+1}^n \sum_{l=1}^m 2^{-k} 2^l \sigma_k \tau_l} \\
 &\quad \times e^{-\pi i \sum_{k,l=m+1}^n 2^{-k} 2^l \sigma_k \tau_l} \\
 &= \mathbf{F}_m(\sigma_{1:m}, \tau_{1:m}) \mathbf{F}_{n-m}(\sigma_{m+1:n}, \tau_{m+1:n}) e^{-2\pi i x_{>m} y_{\leq m}}
 \end{aligned}$$

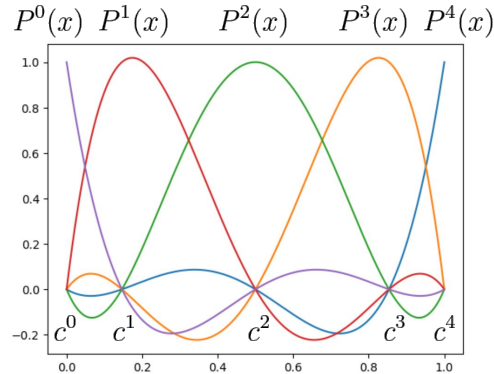

 This part has low-rank decomposition, implying efficient QTT representation!

where we introduced new variables:

$$x_{>m} = \sum_{k=m+1}^n 2^{m-k} \sigma_k \in [0, 1], \quad y_{\leq m} = \sum_{l=1}^m 2^{l-m-1} \tau_l \in [0, 1]$$

Step 2: Low-rank decomposition by Lagrange interpolation

Lagrange interpolation on shifted Chebyshev-Lobatto grids:



$$f(x) \approx \sum_{\alpha=0}^K f(c^\alpha) P^\alpha(x)$$

$$c^\alpha = \frac{1}{2} \left[\cos \left(\frac{\pi(K - \alpha)}{K} \right) + 1 \right]$$

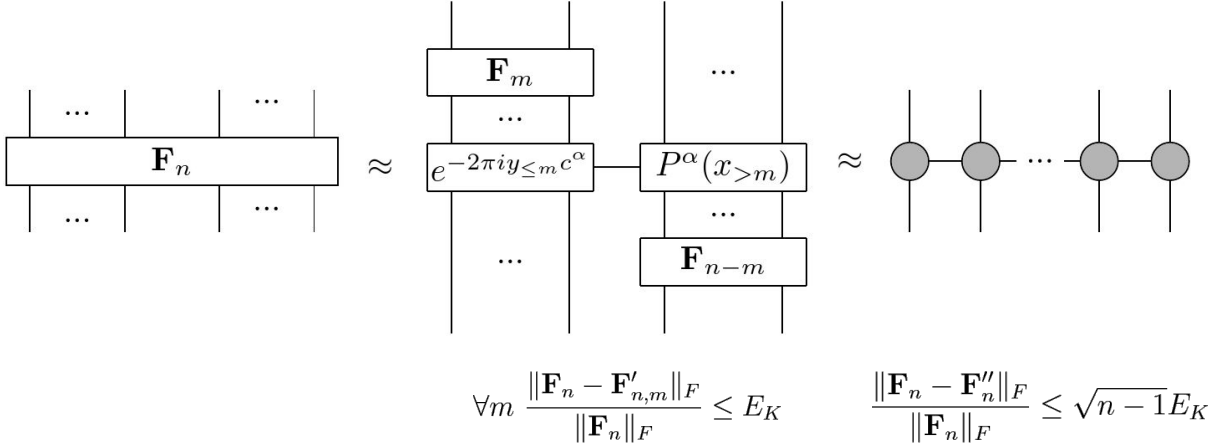
Entry-wise error from interpolation for our case:

Independent of n, m
i.e. true for all sizes & all bonds

$$\|e^{-2\pi i x_{>m} y_{\leq m}} - \sum_{\alpha=0}^K e^{-2\pi i y_{\leq m} c^\alpha} P^\alpha(x_{>m})\|_\infty \leq \frac{4 \left(\frac{\pi}{2}\right)^{K+1} e^K K^{-K}}{K - \frac{\pi}{2}} =: E_K$$

Step 3: QTT with bounded Frobenius norm error

Bounded Frobenius norm error for every bond \rightarrow there exists QTT with controlled Frobenius norm error [1]

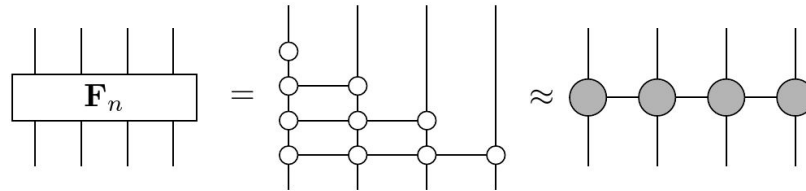


Qualitatively same results in [2], except singular values are directly bounded

[1] I. V. Oseledets, *Tensor-Train Decomposition*, SIAM Journal on Scientific Computing 2011 33:5, 2295-2317
 [2] JC, E.M. Stoudenmire, and Steven R. White, *Quantum Fourier Transform Has Small Entanglement*, PRX Quantum (2023) 4, 040318

How to construct such QTT?

1. TT-SVD (i.e. apply SVD sequentially on every bond) [1]
 - Gives QTT with Frobenius norm error $\sqrt{n-1}E_K$ (quasi-optimal)
 - Exponential time to construct
2. Expand DFT as $n \times n$ 2D tensor network and approximately contract [2]
 - Hard to analyze error; accurate in practice
 - $O(n^2 K^3)$ time to construct



3. Recursive interpolation [3]

- Controlled entry-wise error, but worse than optimal
- Entries have analytical expressions, in principle no need to construct

[1] I. V. Oseledets, *Tensor-Train Decomposition*, SIAM Journal on Scientific Computing 2011 33:5, 2295-2317

[2] JC, E.M. Stoudenmire, and Steven R. White, *Quantum Fourier Transform Has Small Entanglement*, PRX Quantum (2023) 4, 040318

[3] JC, Michael Lindsey, *Direct interpolative construction of the discrete Fourier transform as a matrix product operator*, arXiv:2404.03182 (2024)

Analytical construction by recursive interpolation

Define the m-th interpolative tensor:

$$\begin{array}{c} \sigma_1 \quad \dots \quad \sigma_m \\ | \quad \quad \quad | \\ \boxed{F_m^\alpha(\sigma_{1:m}, \tau_{1:m})} \\ | \quad \quad \quad | \\ \tau_1 \quad \dots \quad \tau_m \end{array} \xrightarrow{\alpha} = \begin{array}{c} \sigma_1 \quad \dots \quad \sigma_m \\ | \quad \quad \quad | \\ \boxed{\mathbf{F}_m} \\ | \quad \quad \quad | \\ \boxed{e^{-2\pi i y_{\leq m} c^\alpha}} \\ | \quad \quad \quad | \\ \tau_1 \quad \dots \quad \tau_m \end{array} \xrightarrow{\alpha} = e^{-\pi i \sum_{k,l=1}^m 2^{-k} 2^l \sigma_k \tau_l} e^{-2\pi i y_{\leq m} c^\alpha}$$

Interpolate the (m+1)-th interpolative tensor with the m-th one:

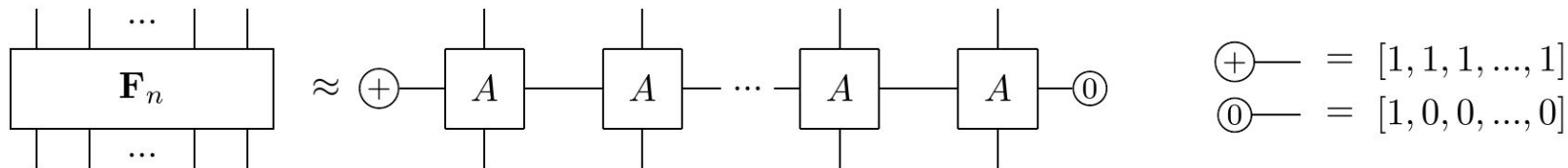
$$\begin{array}{c} \sigma_1 \quad \dots \quad \sigma_m \quad \sigma_{m+1} \\ | \quad \quad \quad | \quad | \\ \boxed{F_{m+1}^\alpha(\sigma_{1:m+1}, \tau_{1:m+1})} \\ | \quad \quad \quad | \quad | \\ \tau_1 \quad \dots \quad \tau_m \quad \tau_{m+1} \end{array} \xrightarrow{\beta} \approx \begin{array}{c} \sigma_1 \quad \dots \quad \sigma_m \quad \sigma_{m+1} \\ | \quad \quad \quad | \quad | \\ \boxed{F_m^\alpha(\sigma_{1:m}, \tau_{1:m})} \\ | \quad \quad \quad | \quad | \\ \tau_1 \quad \dots \quad \tau_m \quad \tau_{m+1} \end{array} \xrightarrow{\alpha} \boxed{A} \xrightarrow{\beta}$$

Solution:

$$\begin{aligned} & A^{\alpha\beta}(\sigma, \tau) \\ &= P^\alpha \left(\frac{\sigma + c^\beta}{2} \right) e^{-\pi i (\sigma + c^\beta) \tau} \end{aligned}$$

Analytical construction by recursive interpolation

Gives QTT with translation-invariant sites and closed boundaries



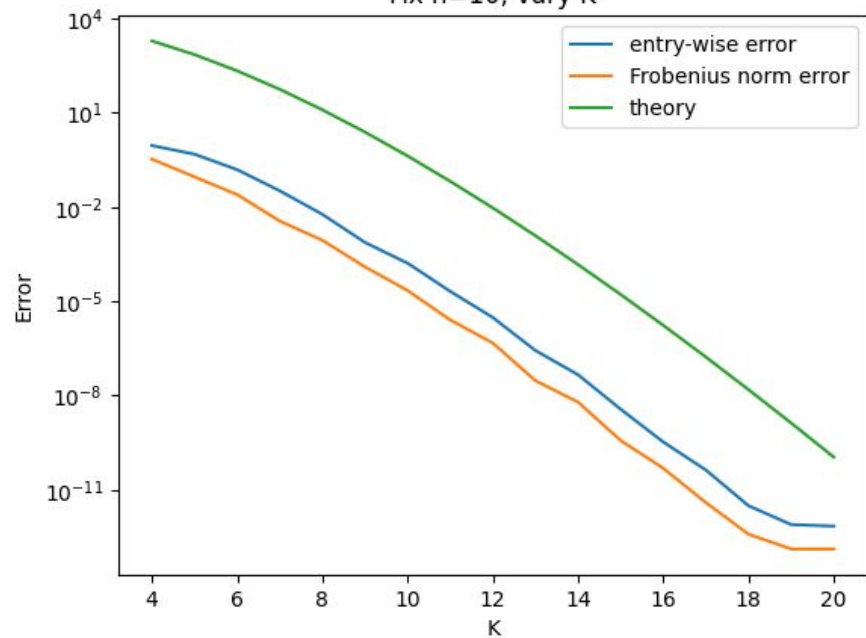
Entry-wise error bound:

$$\max_{s,t} |\mathbf{F}_n(s,t) - \mathbf{F}_n''(s,t)| \leq \frac{\Lambda_K^{n-1} - 1}{\Lambda_K - 1} E_K \quad \text{Worse than optimal}$$

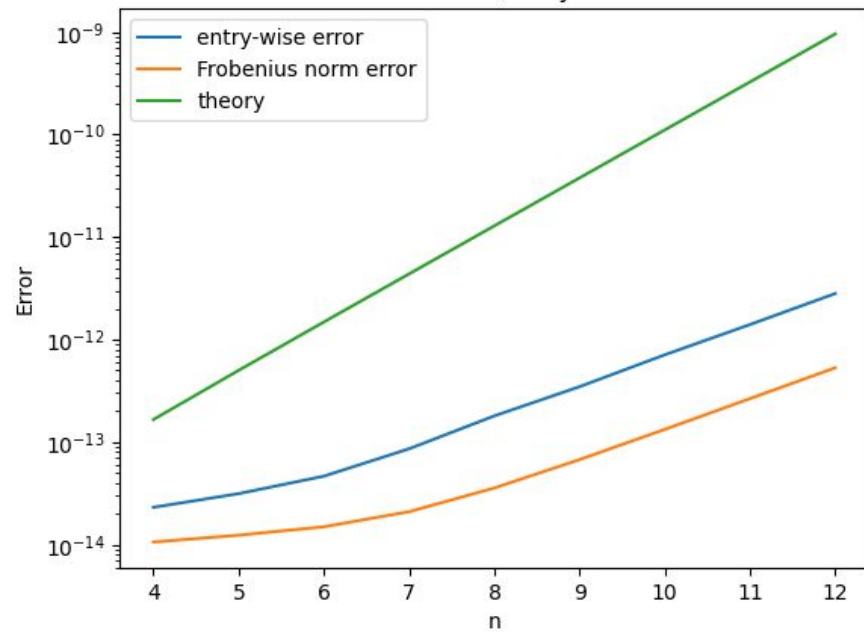
$$\Lambda_K \leq 1 + \frac{2}{\pi} \log(K + 1) \quad \text{Lebesgue constant for Chebyshev-Lobatto grids}$$

Numerics

Fix $n=10$, vary K



Fix $K=20$, vary n



Conclusion

- We proved that DFT can be represented as a low-rank QTT
- We gave an analytical construction through interpolative decomposition

Discussion

- Tighter bounds?
- Generalize to other operators?
- Applications of the analytical construction?

Acknowledgement

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Thanks for discussion: Sandeep Sharma, Garnet Chan