

Tensorizing Feynman diagrams

Quantics tensor trains and quantics tensor cross interpolation

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FY2023-2025 MEXT -KAKENHI- Grant-in-Aid for Transformative Research Areas (B)
“Computational materials science based on quantum-classical hybrid algorithms”



Collaborators

📌 HS *et al.*, PRX **13**, 021015 (2023)

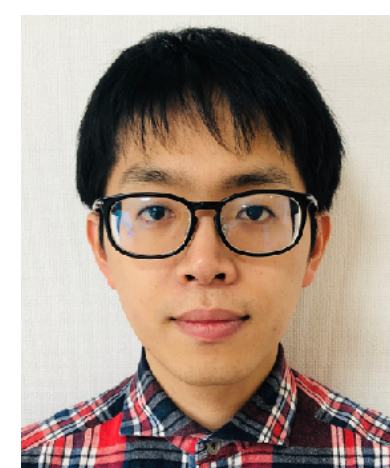


M. Wallerberger

TU Wien

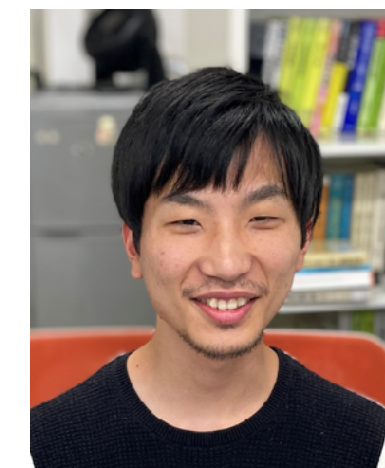


A. Kauch



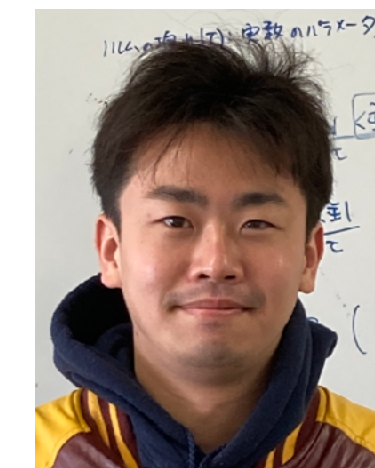
Y. Murakami

Riken



K. Nogaki

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R. Sakurai

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P. Werner

Fribourg Univ.

📌 M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, **HS**, X. Waintal, arXiv:2303.11819



M. K. Ritter

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J. von Delft



Y. N. Fernández

CEA Grenoble

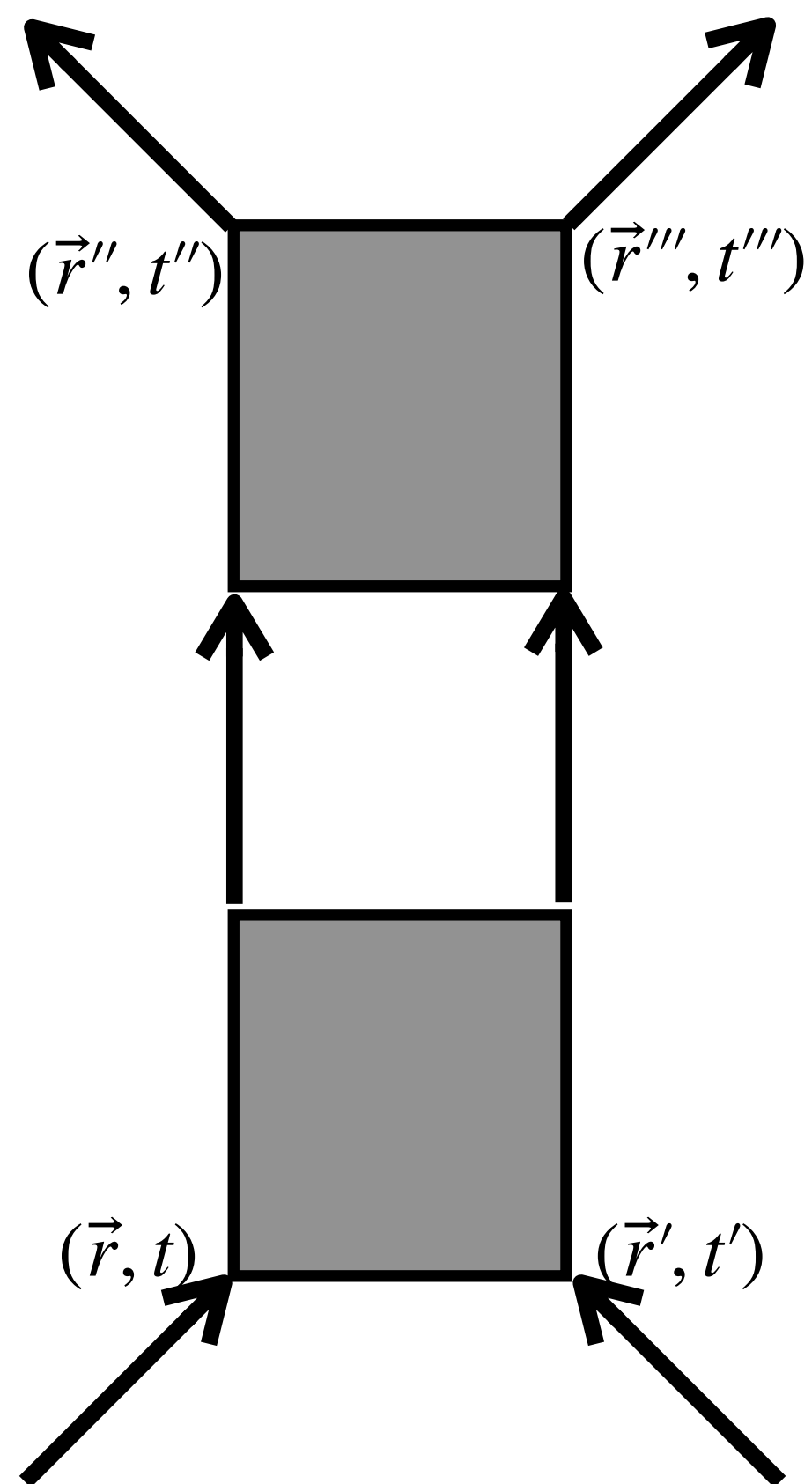


X. Waintal

Special thanks to E. Miles Stoudenmire (CCQ) for his help in implementing our code with ITensors.jl



Tensors for all field theories!



“Quantics” allows

- Compression: Coexisting (exponentially) different length scales
- Computation: Integration, Fourier transform, convolution

Benchmarks: Solving Dyson/Bethe equations, BZ integration...

Beyond condensed matter physics!

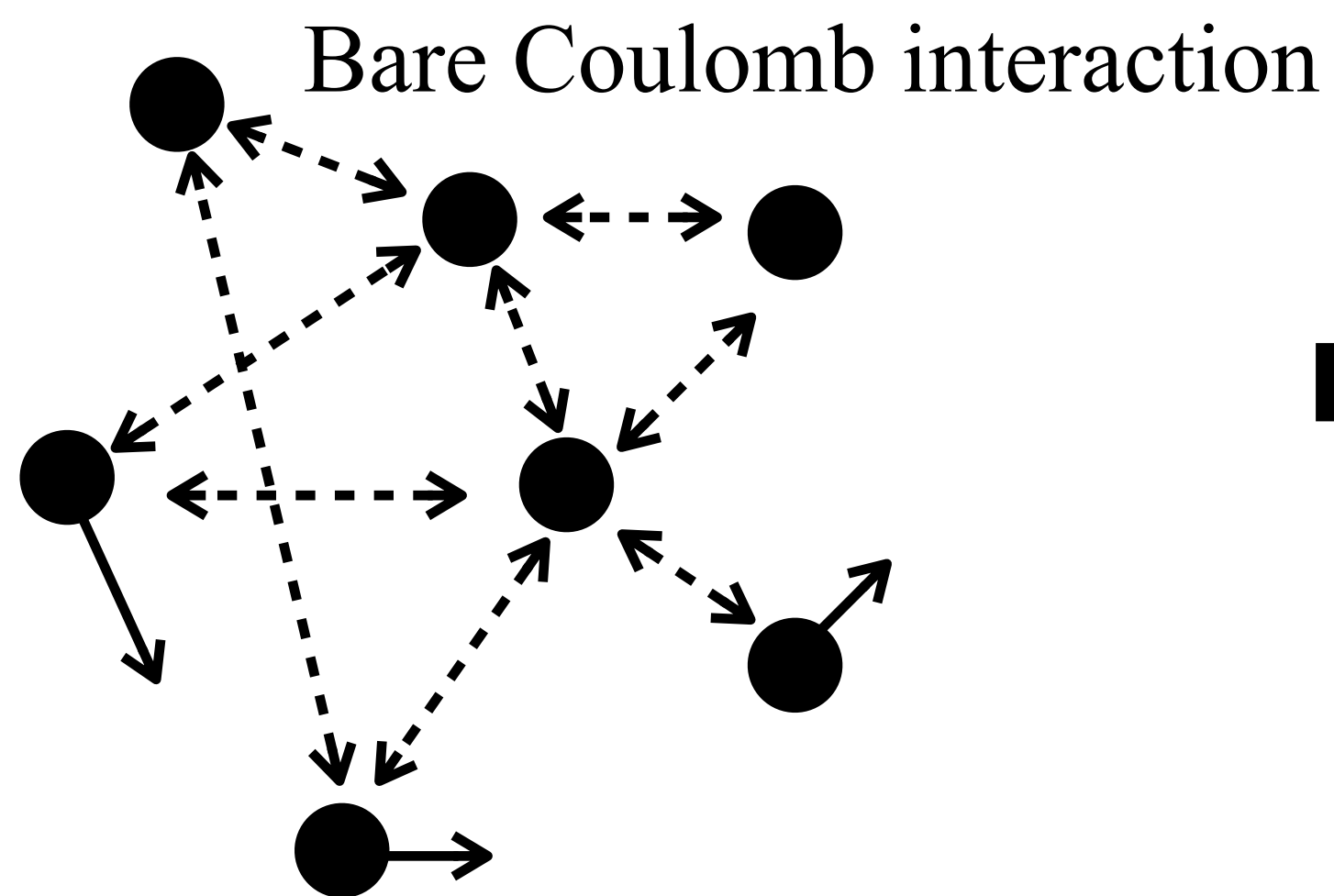
Outline

- Introduction
- Part I Quantics tensor train
HS et al., PRX **13**, 021015 (2023)
- Part II Quantics tensor cross interpolation
M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, **HS**, X. Waintal, arXiv:2303.11819
- Outlook & summary

Quantum field theories

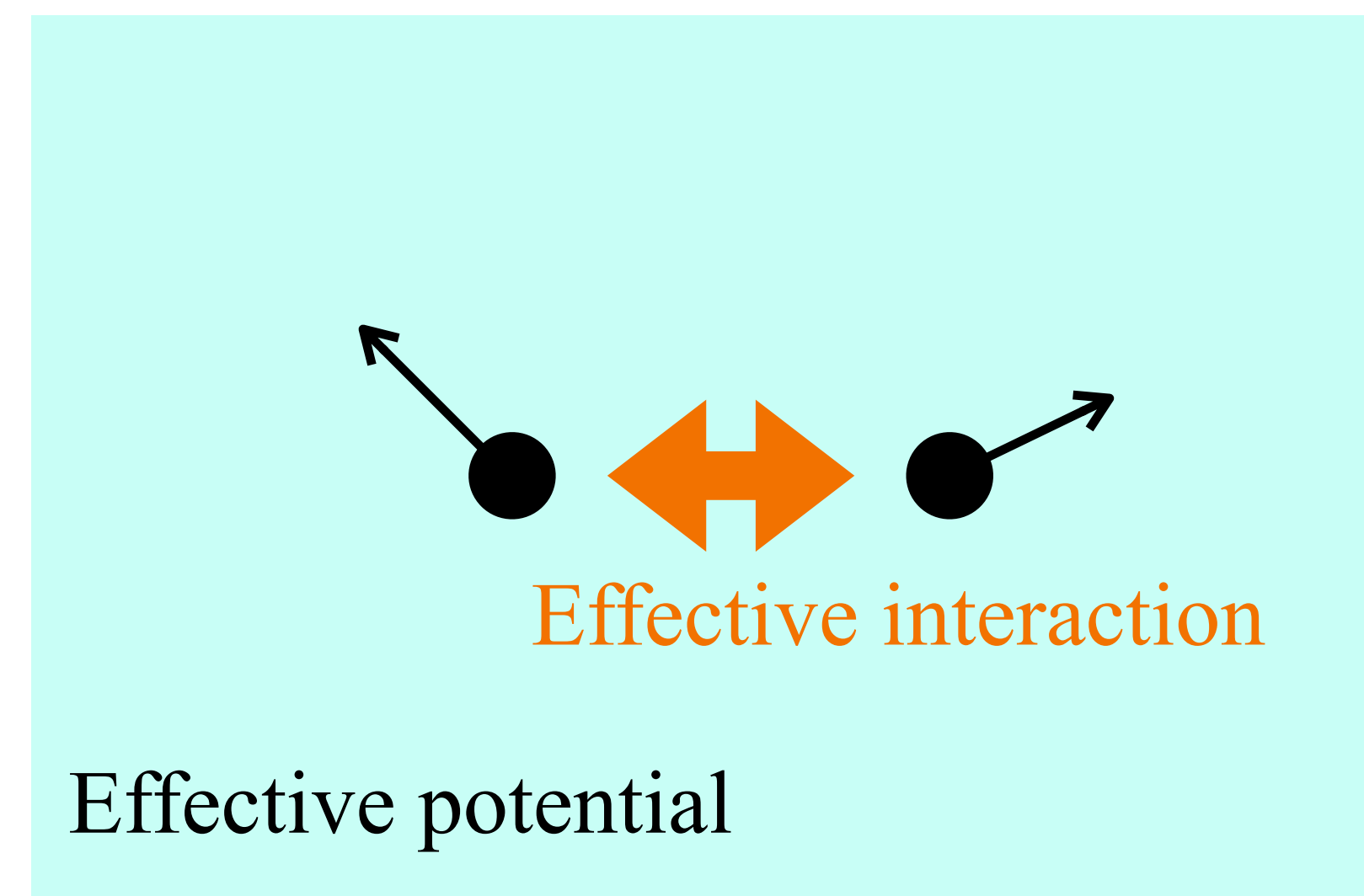
with a focus on many-body electronic systems

Many-particle interacting system



Exponential growth in computational cost

Effective few-particle system



DFT (Kohm-Sham), diagrammatic many-body theories, DMFT, *etc.*

Grand challenges

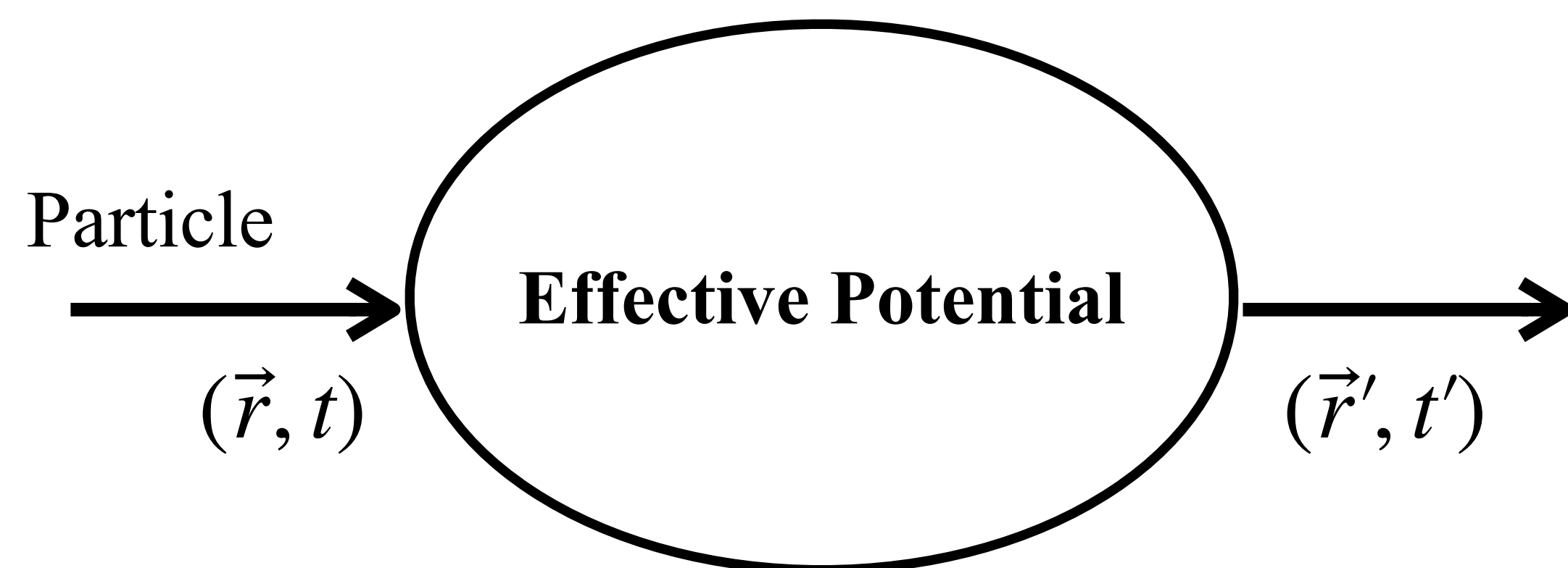
- More accurate mapping
- Still very expensive to solve effective few-particle systems

Building block of quantum field theory

Correlation functions in a high-dimensional space-time domain

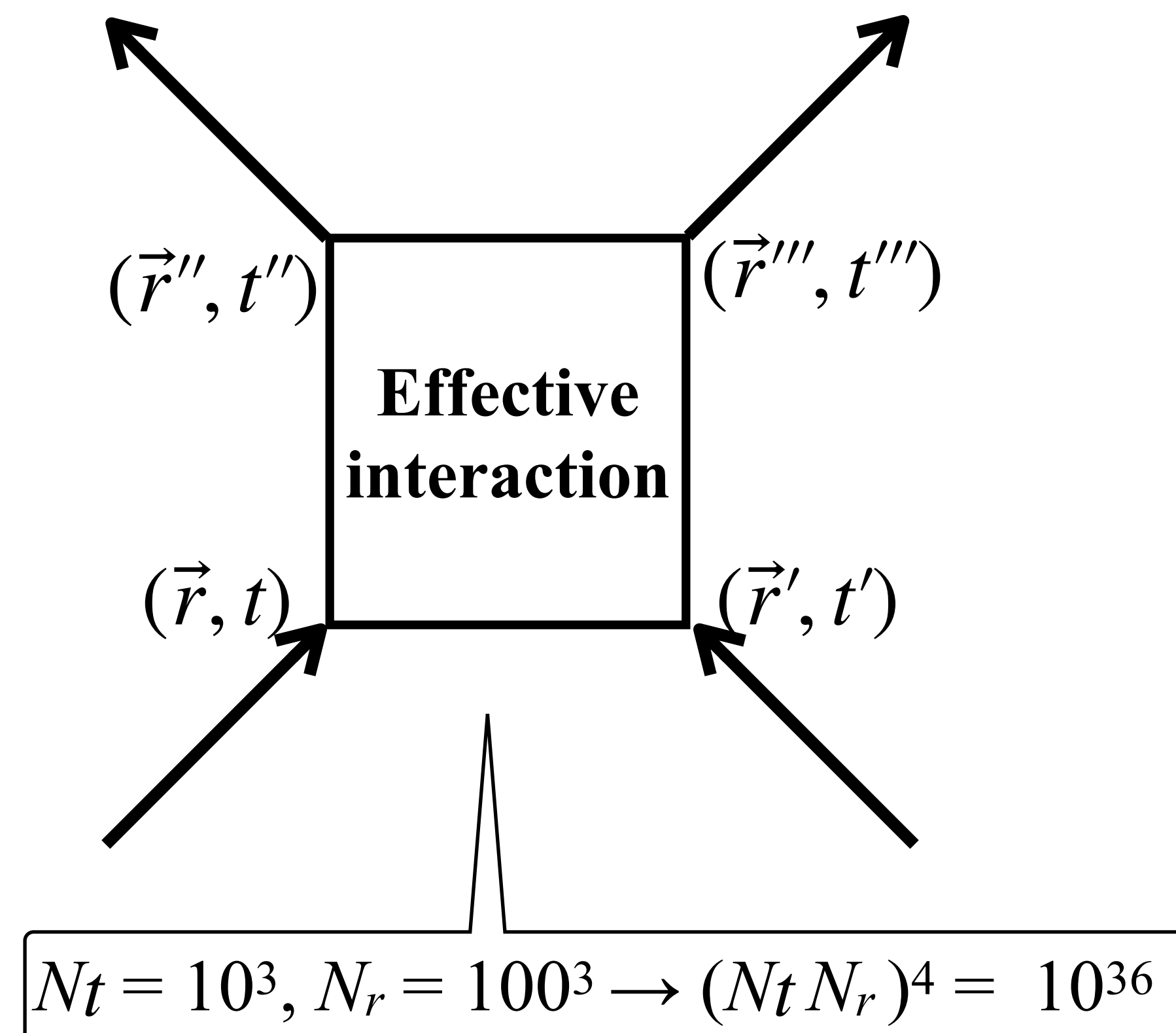
One-particle (1P) level

Symmetry breaking



Two-particle (2P) level

Susceptibility, beyond mean field



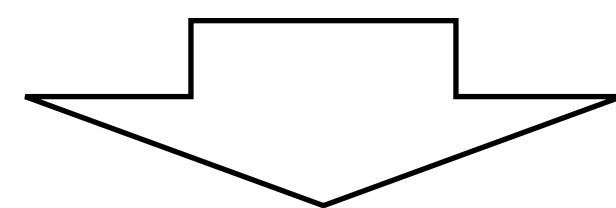
Matsubara-frequency domain

imaginary time/Euclidean time

Required ability to handle a wide range of energy scales $> 10^4$

From band width (>100 eV) to low temperature (1K \sim 0.1 meV)

Prior knowledge $G(\tau)$ is related to $\rho(\omega)$ through ill-posed analytic continuation kernel



$$G(\tau) = \int_0^\beta d\omega K(\tau, \omega) \rho(\omega)$$

- Intermediate representation + sparse sampling

HS *et al.*, PRB **96**, 035147 (2017)

J. Li *et al.*, HS, PRB **101**, 035144 (2020)

HS *et al.*, SciPost Phys. Lect. Notes 63 (2022)

- Minmax method

M. Kaltak and G. Kresse, PRB **101**, 205145 (2020)

- Discrete Lehmann Representation

J. Kaye *et al.*, PRB **105**, 235115 (2022)

- *Ab initio* Migdal-Eliashberg calculation
T. Wang, ..., HS, ... R. Arita, PRB **102**, 134503 (2020)
- Multi-orbital FLEX for unconventional superconductivity
N. Witt *et al.*, PRB **103**, 205148 (2021)
- *Ab initio* self-energy embedding for transition metal oxides
S. Isakov *et al.*, PRB **102**, 085105 (2020)

The idea does not generalize to other domains, e.g., real-time, momentum space...

Other domains

Multi Matsubara domain

Overcomplete basis based on analytic continuation kernel

HS *et al.*, PRB **97**, 205111 (2018), HS *et al.*, SciPost Phys. **8**, 012 (2020), M. Wallerberger, HS, A. Kauch, PRR **3**, 033168 (2021), S.-S. B. Lee *et al.*, PRX **11**, 041007 (2021), F. B. Kugler *et al.*, PRX **11**, 041006 (2021)

Computation on the overcomplete basis is cumbersome.

Real-time (non-equilibrium) domain

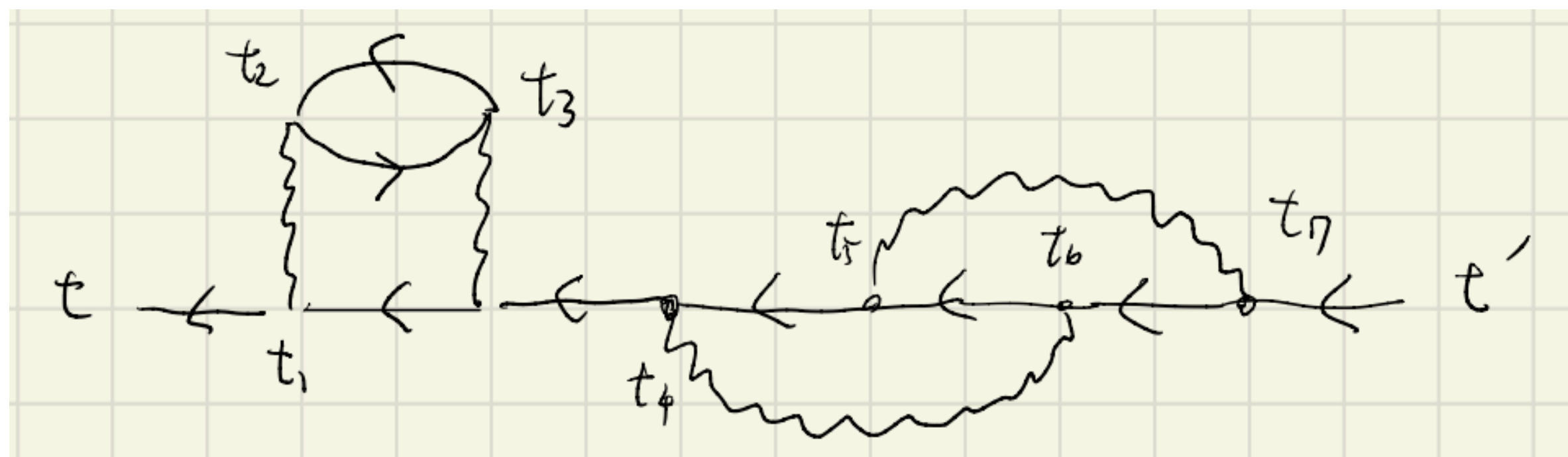
Hierarchical low-rank compression J. Kaye, Denis Golež, SciPost Phys. **10**, 091 (2021)

Multi momentum domain

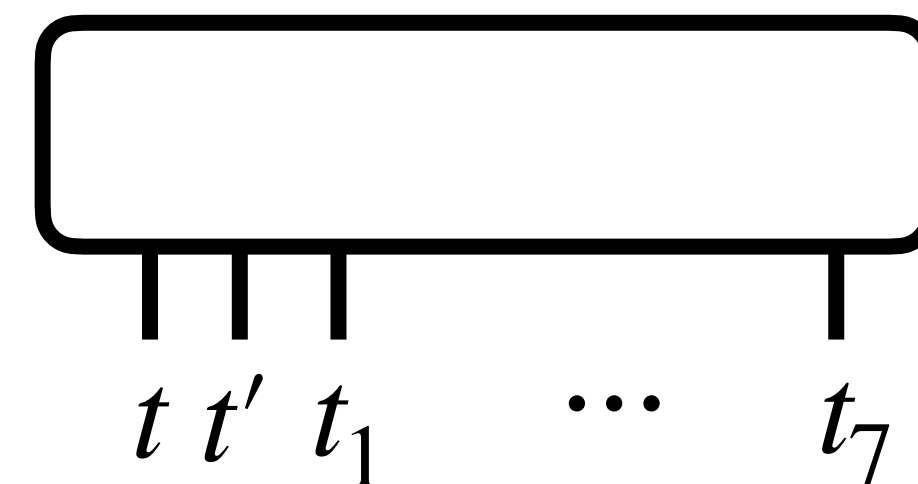
Truncated form-factor basis C. J. Eckhardt *et al.*, PRB **98**, 075143 (2018), C. J. Eckhardt *et al.*, PRB **101**, 155104 (2020)

General compact bases are still under active development.

Feynman diagram/correlation function is tensor



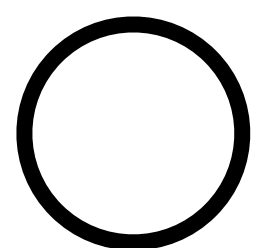
$$= \int dt_1 \cdots dt_7 F(t, t'; t_1, \cdots, t_7) = G(t, t')$$



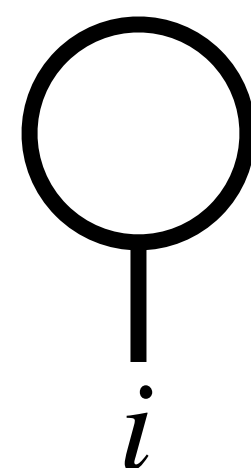
Any *compact* tensor network representation?

Warm-up: Tensor diagram notation

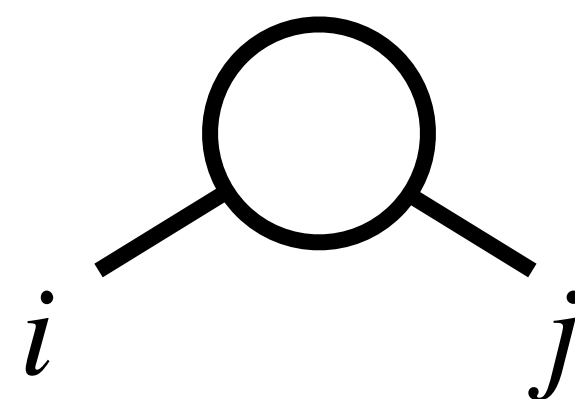
Scalar

 a 

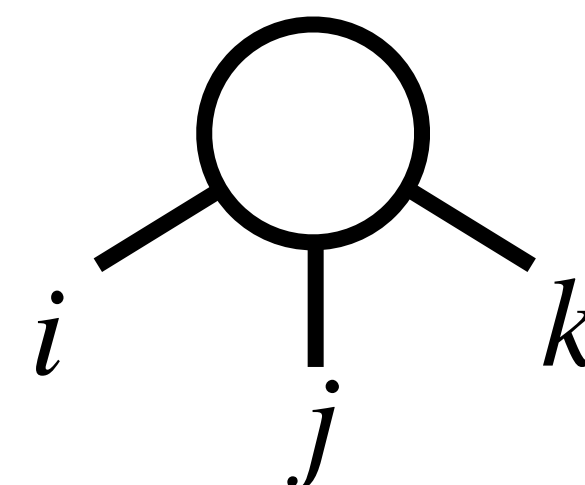
Vector

 a_i 

Matrix

 a_{ij} 

Three-way tensor

 a_{ijk} 

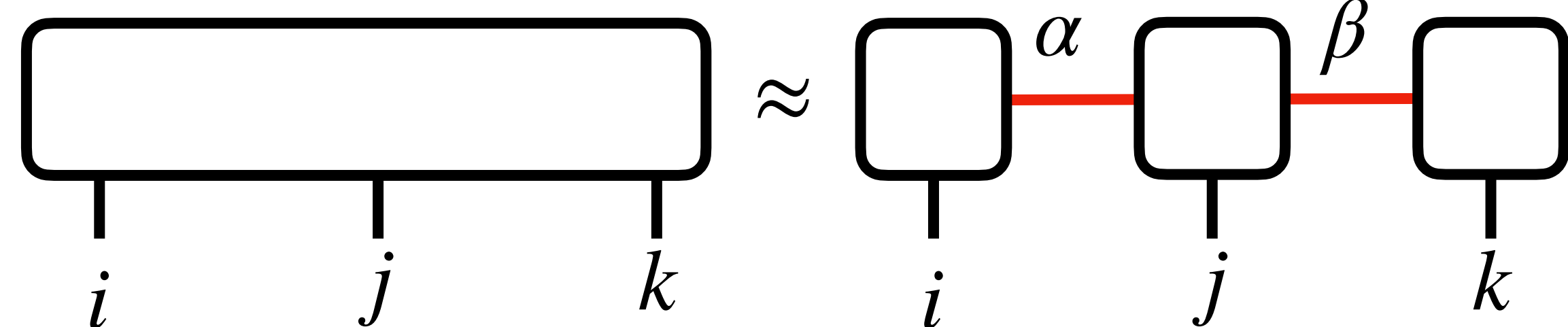
A leg sticking out corresponds to an index.

Contraction

$$\sum_j A_{ij} x_j = \begin{array}{c} i \text{ --- } \boxed{A} \text{ --- } j \text{ --- } \boxed{x} \\ \\ i \text{ --- } \boxed{A} \end{array}$$

Summation over shared indices

Tensor train/matrix product state



Virtual bond

Compression for small bond dimensions!

Singular value decomposition, cross interpolation...

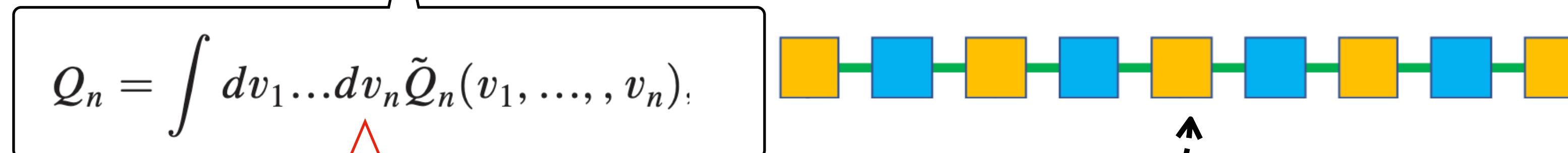
Previous study: Learning Feynman diagram

Y. N. Fernández *et al.*, PRX **12**, 041018 (2022)

Weak-coupling
expansion in real time

$$H = H_0 + UH_{\text{int}} \quad Q(U) = \sum_n Q_n U^n \quad \text{Charge, Green's function, etc.}$$

Single-orbital Anderson
impurity model

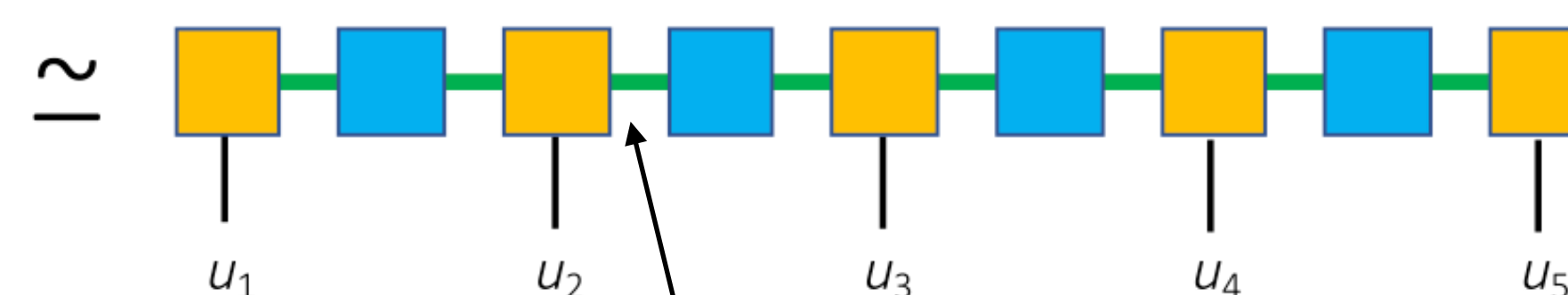
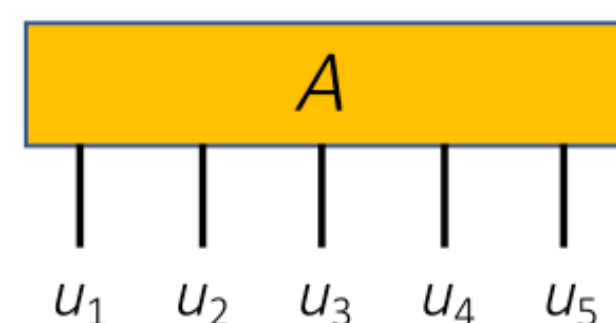


$$Q_n = \int dv_1 \dots dv_n \tilde{Q}_n(v_1, \dots, v_n)$$

Low-rank approximation by Tensor Cross Interpolation (TCI)

$$\tilde{Q}_n(v_1, \dots, v_n) \approx M_1(v_1) \cdots M_n(v_n) \quad Q_n \approx \left(\int dv_1 M_1(v_1) \right) \cdots \left(\int dv_n M_n(v_n) \right)$$

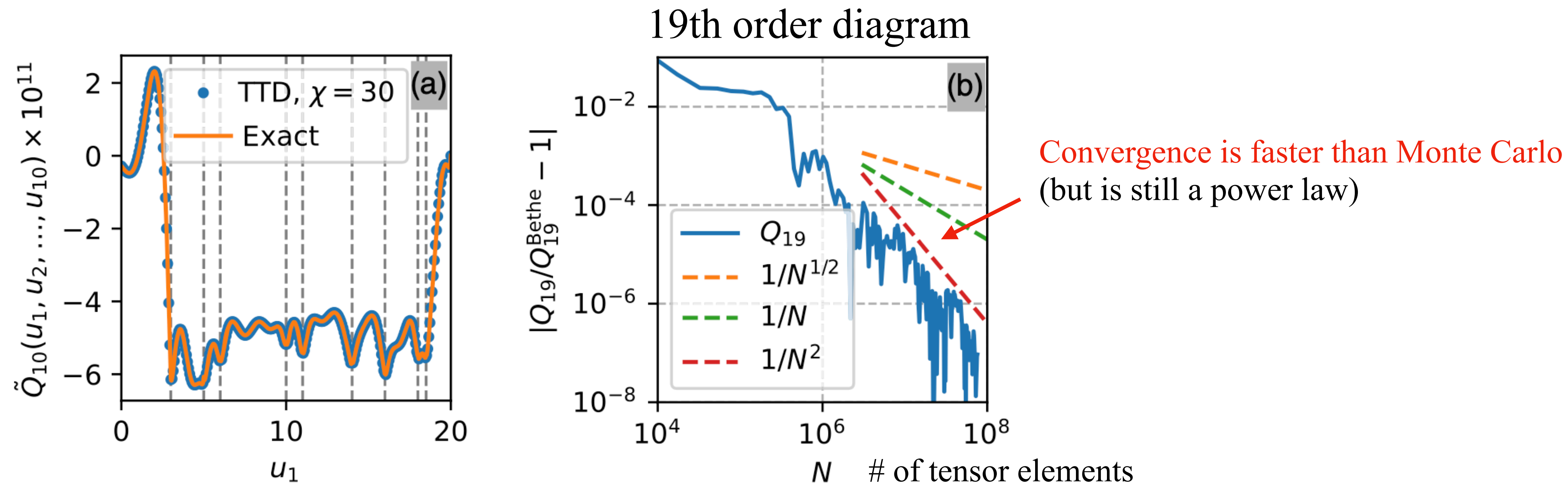
TCI



Non-trivial: Bond dimensions are small?

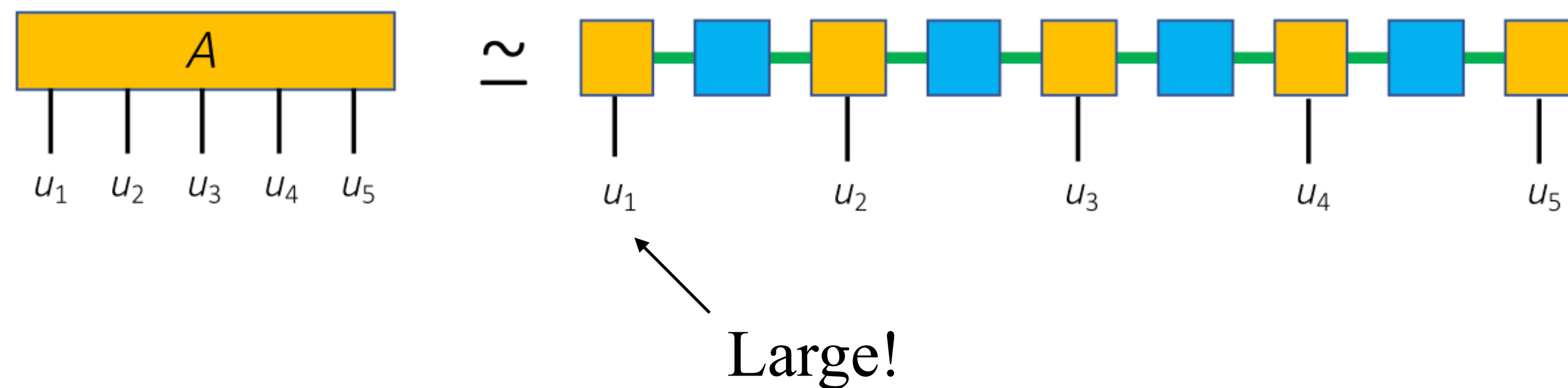
Previous study: Learning Feynman diagram

Y. N. Fernández *et al.*, PRX **12**, 041018 (2022)

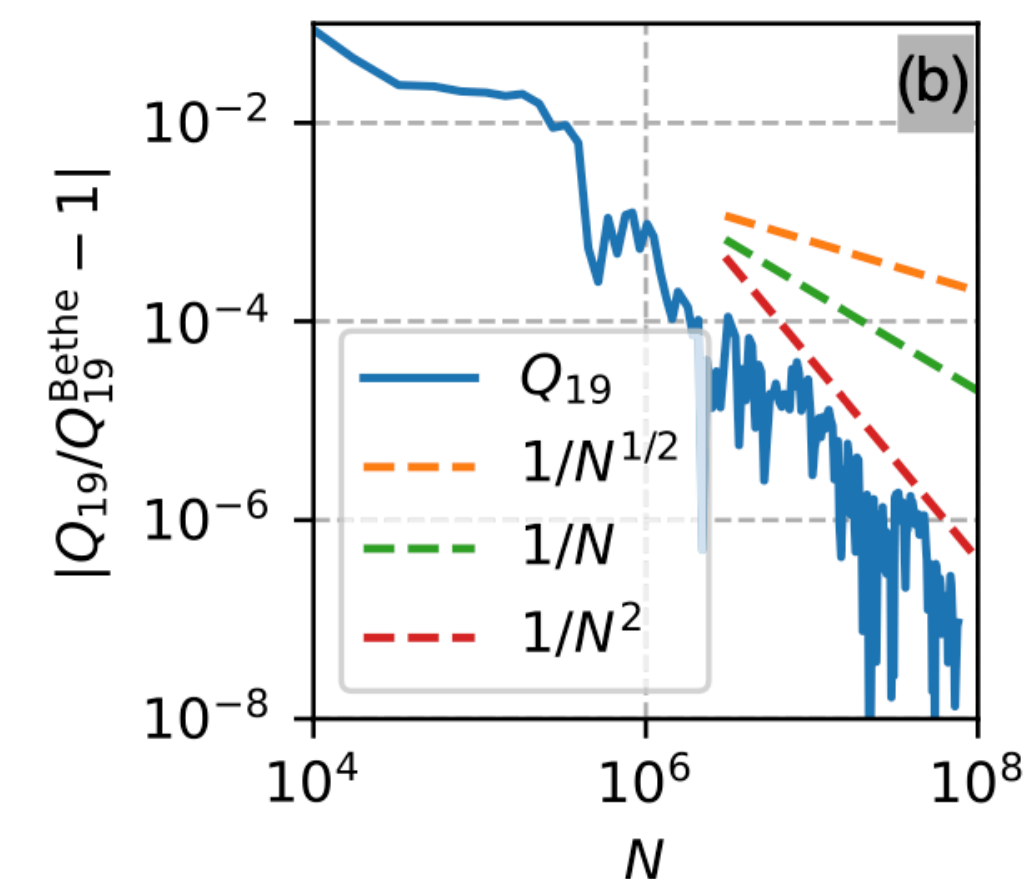


Sign problem free, replacement for quantum Monte Carlo
(in some cases)

Remaining issues



- How to treat a wide range of coexisting length scales?
- Can we achieve exponential convergence?
- Computation (convolution, Fourier transform...)



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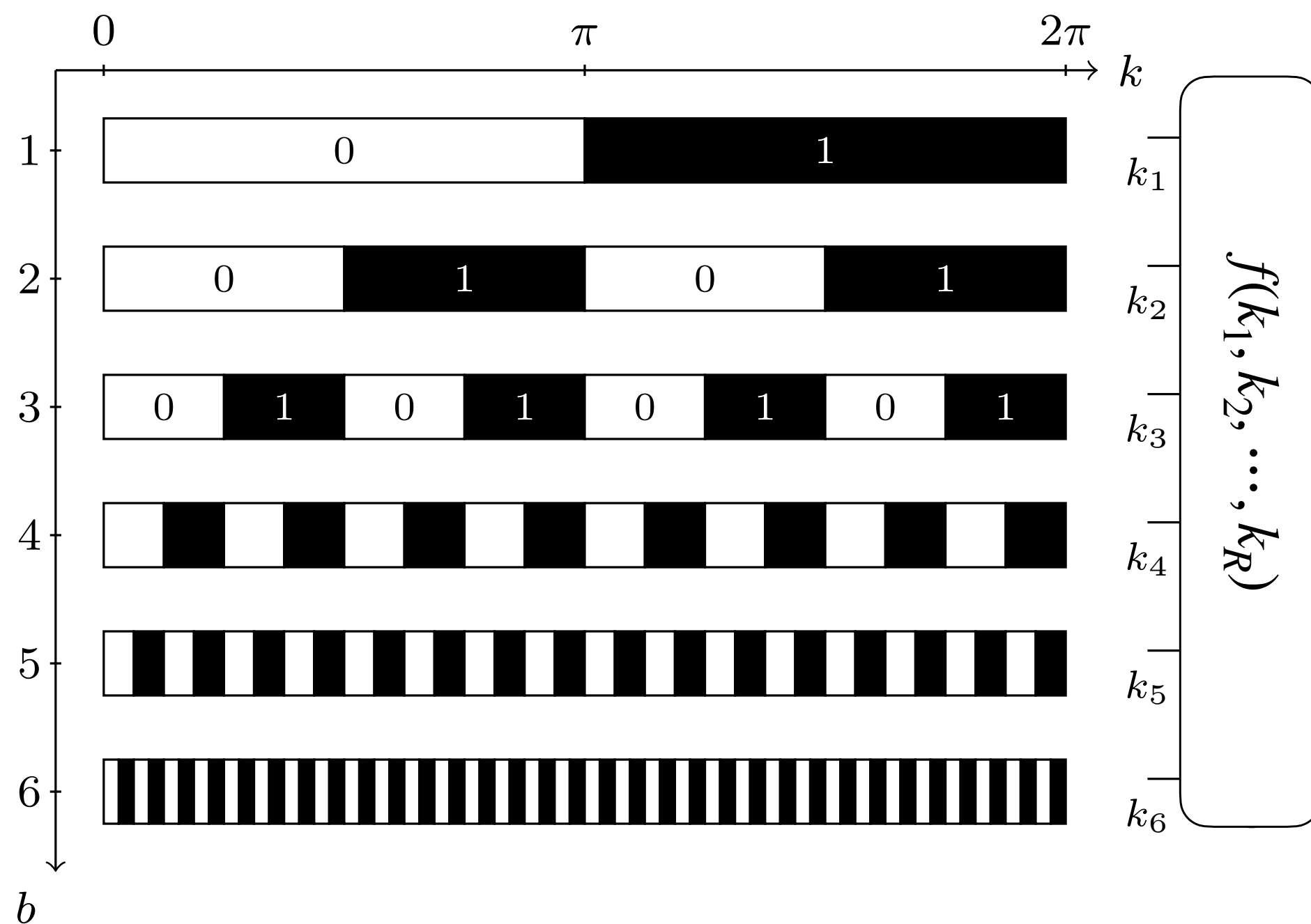
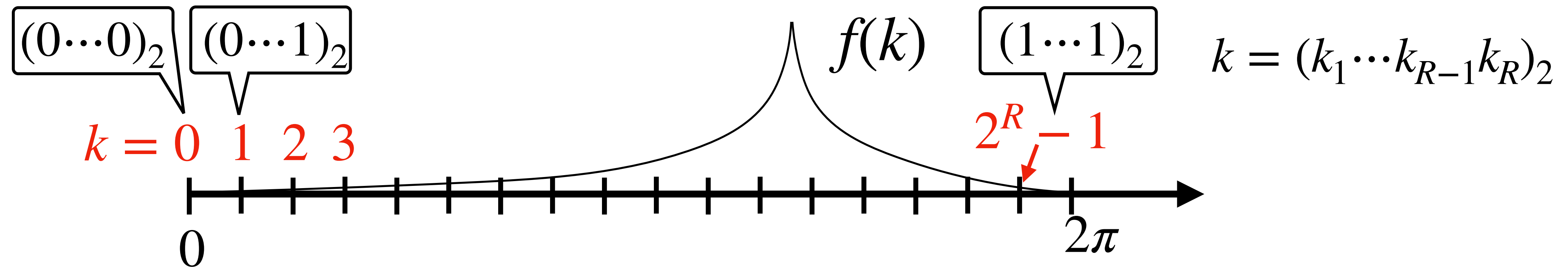
Quantics tensor train (QTT)

I. V. Oseledets, Doklady Math. **80**, 653 (2009)

ExU-YITP 2023/9/28

B. N. Khoromskij, Constr. Approx. **34**, 257 (2011)

For image compression José I. Latorre, arXiv:quant-ph/0510031v1



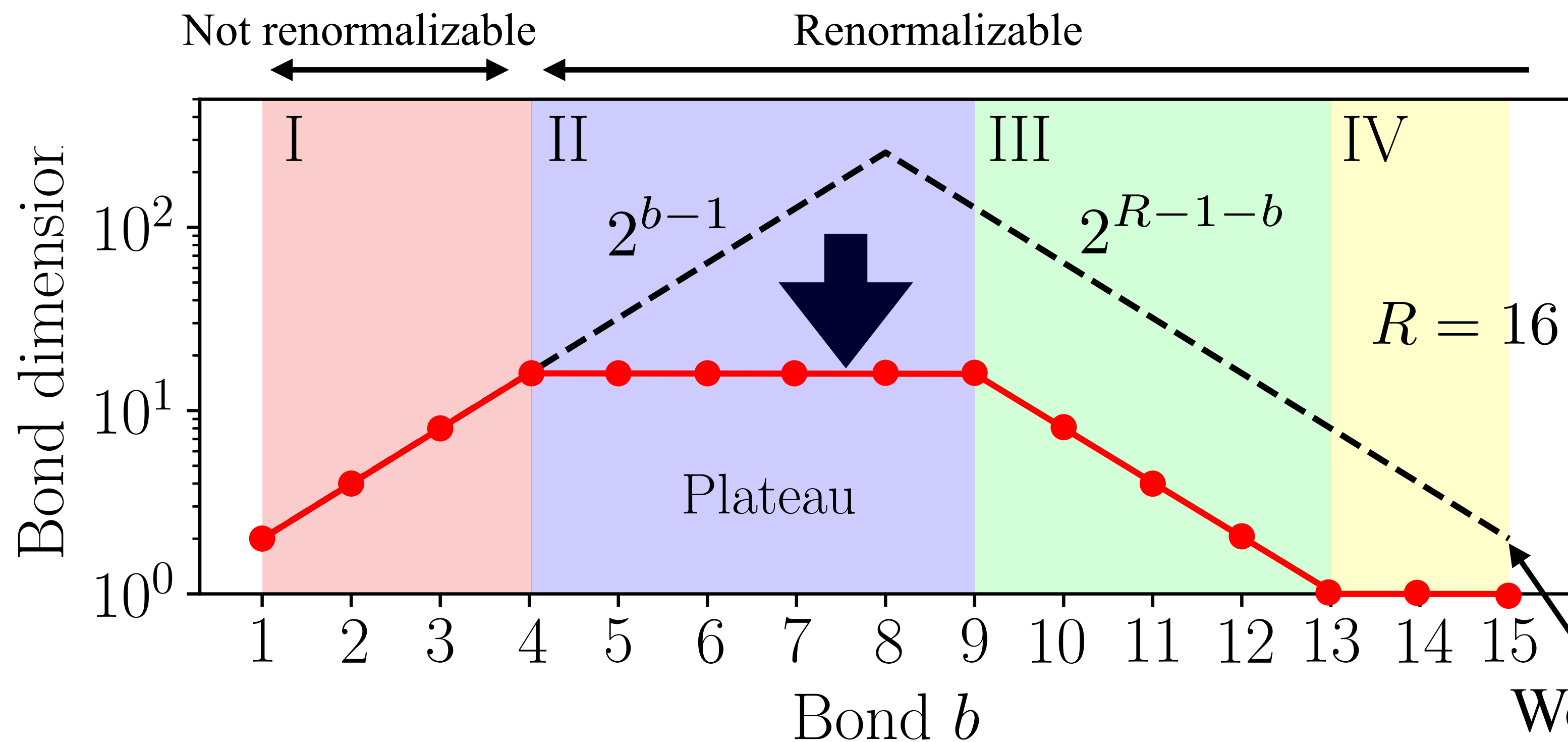
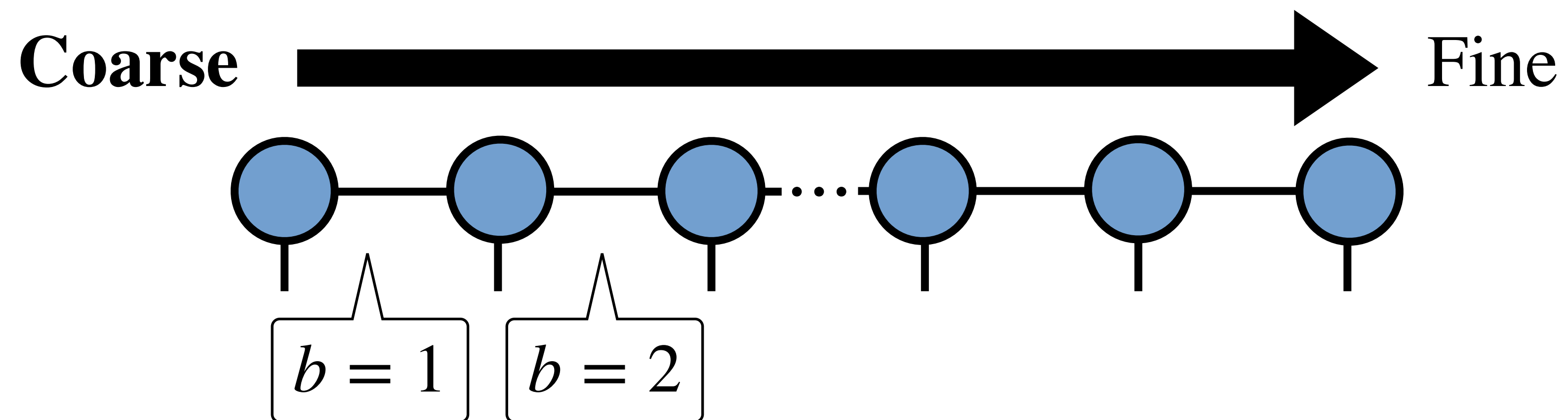
Length-scale separation \rightarrow **Bond dimension** $\ll 2^{R/2}$

$$f(k_1, k_2, \dots, k_R) \approx \sum_{\alpha_1=1}^{D_1} \dots \sum_{\alpha_{R-1}=1}^{D_{R-1}} \hat{F}_{k_1, 1\alpha_1}^{(1)} \hat{F}_{k_1, \alpha_1\alpha_2}^{(2)} \dots \hat{F}_{k_R, \alpha_{R-1}1}^{(R)}$$

Tensor train/Matrix product state

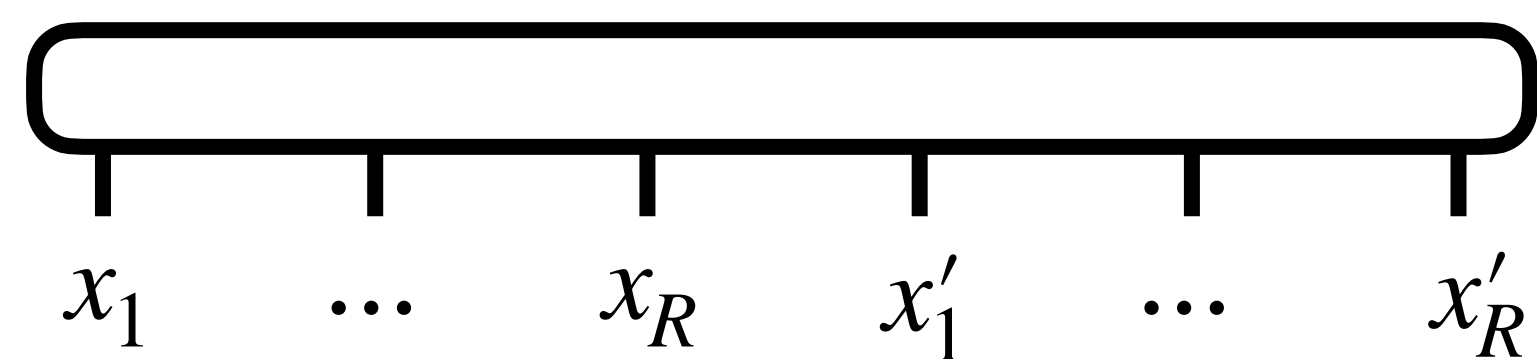
QTT can represent a low-entanglement structure between different length scales!

Length-scale separation



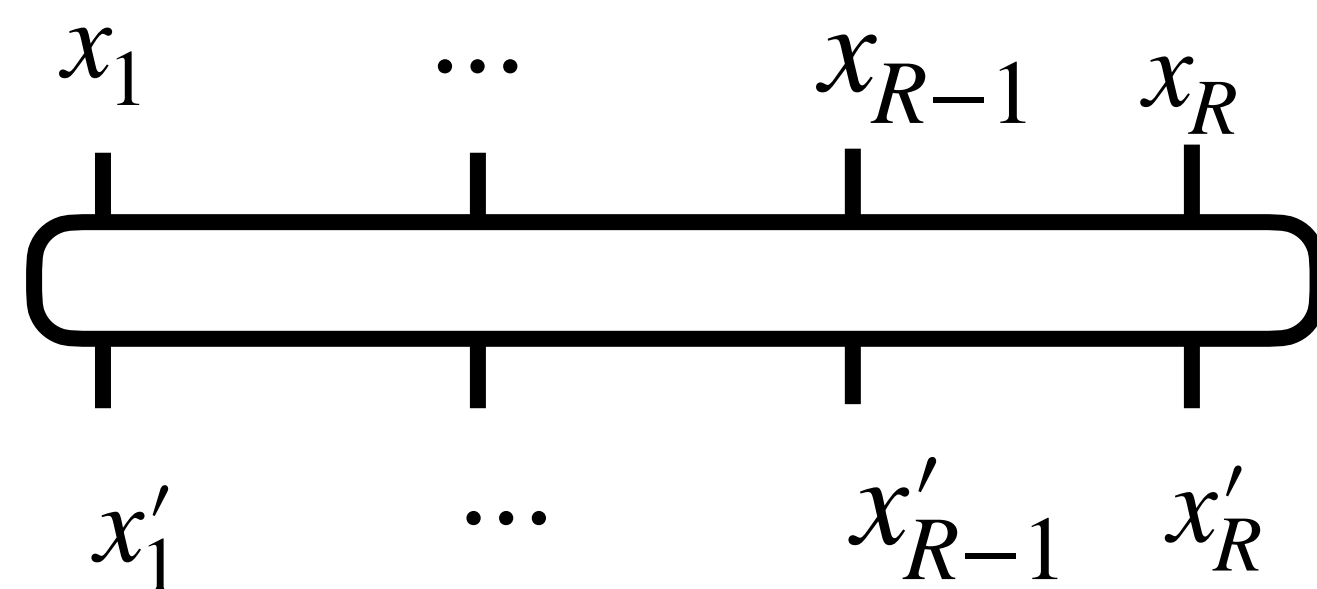
- Exponentially wide range of length scales
- Truncation limits entanglement between length scales.

Multivariate functions

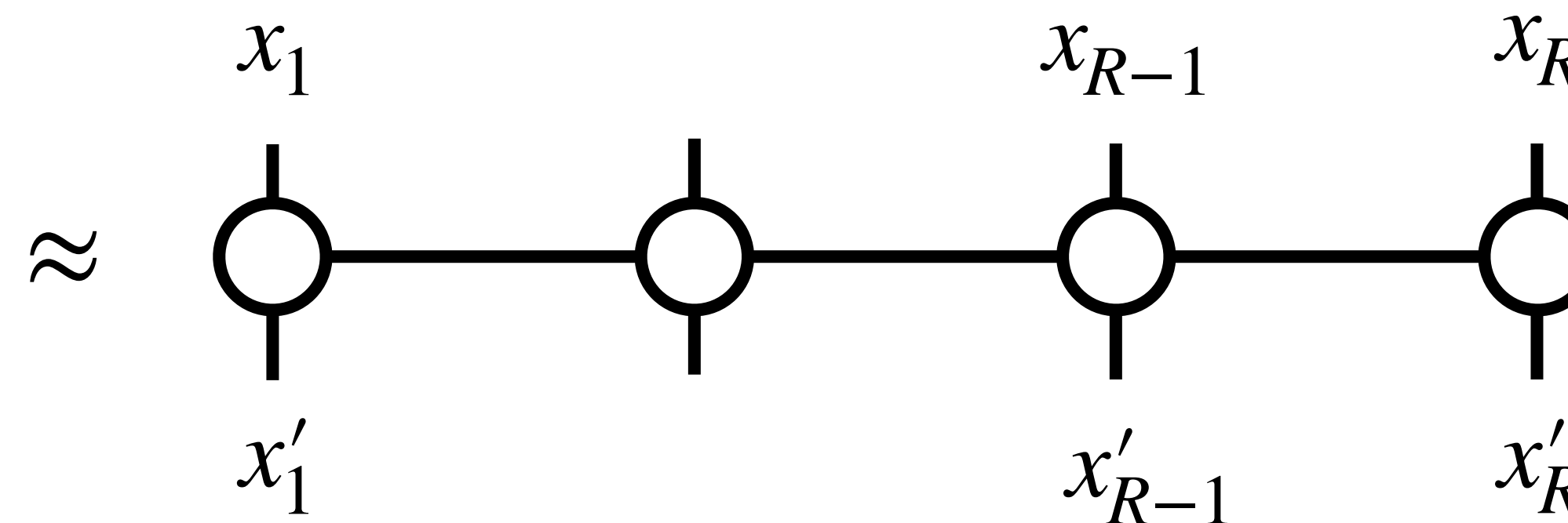
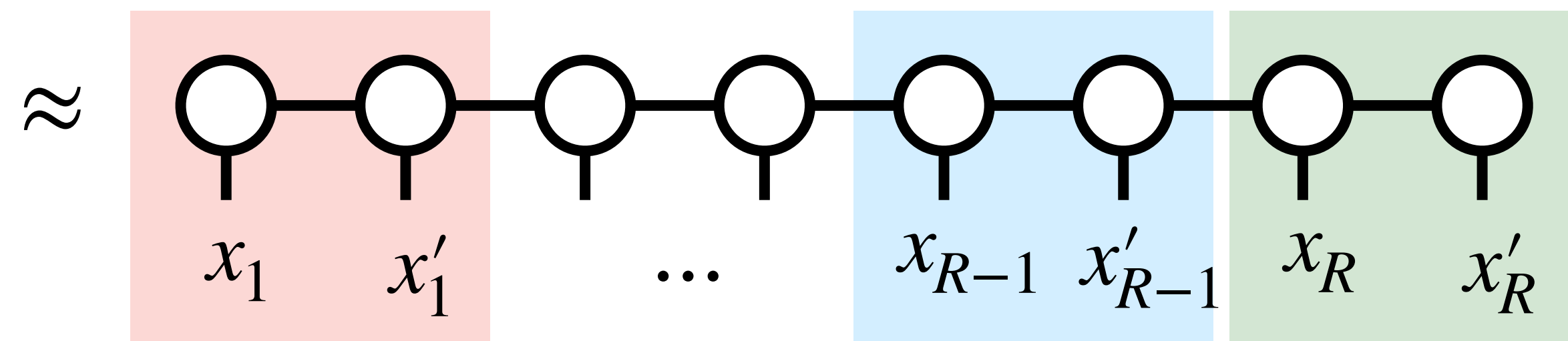
 $f(x, x')$


$$x = (0.x_1x_2\dots x_n\dots)_2 \in [0,1)$$

$$x' = (0.x'_1x'_2\dots x'_n\dots)_2 \in [0,1)$$



Same length scale



“Matrix product operator” (MPO)

Function with trivial QTT representation

Exponential $f(x) = e^{-x} = e^{-x_1/2} e^{-x_2/2^2} \dots e^{-x_n/2^n} \dots$ $D = 1$
 $x = (0.x_1x_2\dots x_n\dots)_2 \in [0,1)$

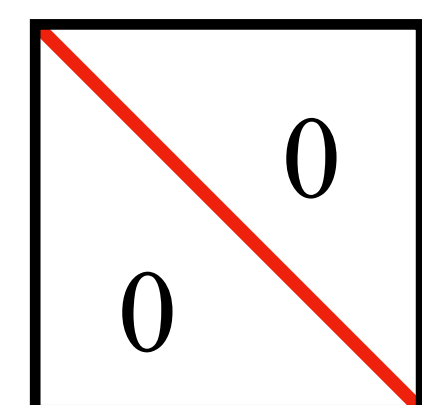
The Sum of N exponential functions can be represented as a QTT of rank at most N .
 \therefore Bond dimensions are added when MPSs are added.

Polynomial $D \leq 1 + p$

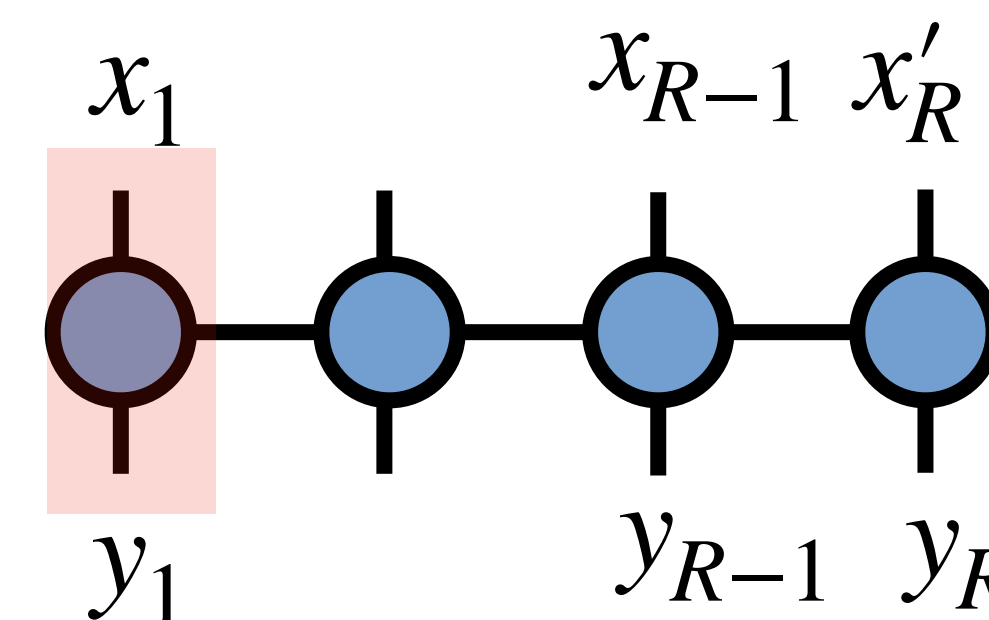
M. Ali and A. Nouy, “A. Approximation Theory of Tree Tensor Networks: Tensorized Univariate Functions”, Constr Approx (2023)

Identity matrix

$f(x, y) = \delta_{x,y} = \delta_{x_1,y_1} \delta_{x_2,y_2} \dots$ $D = 1$



$= \square \otimes \square \otimes \square \otimes \square \otimes \square \otimes \dots$

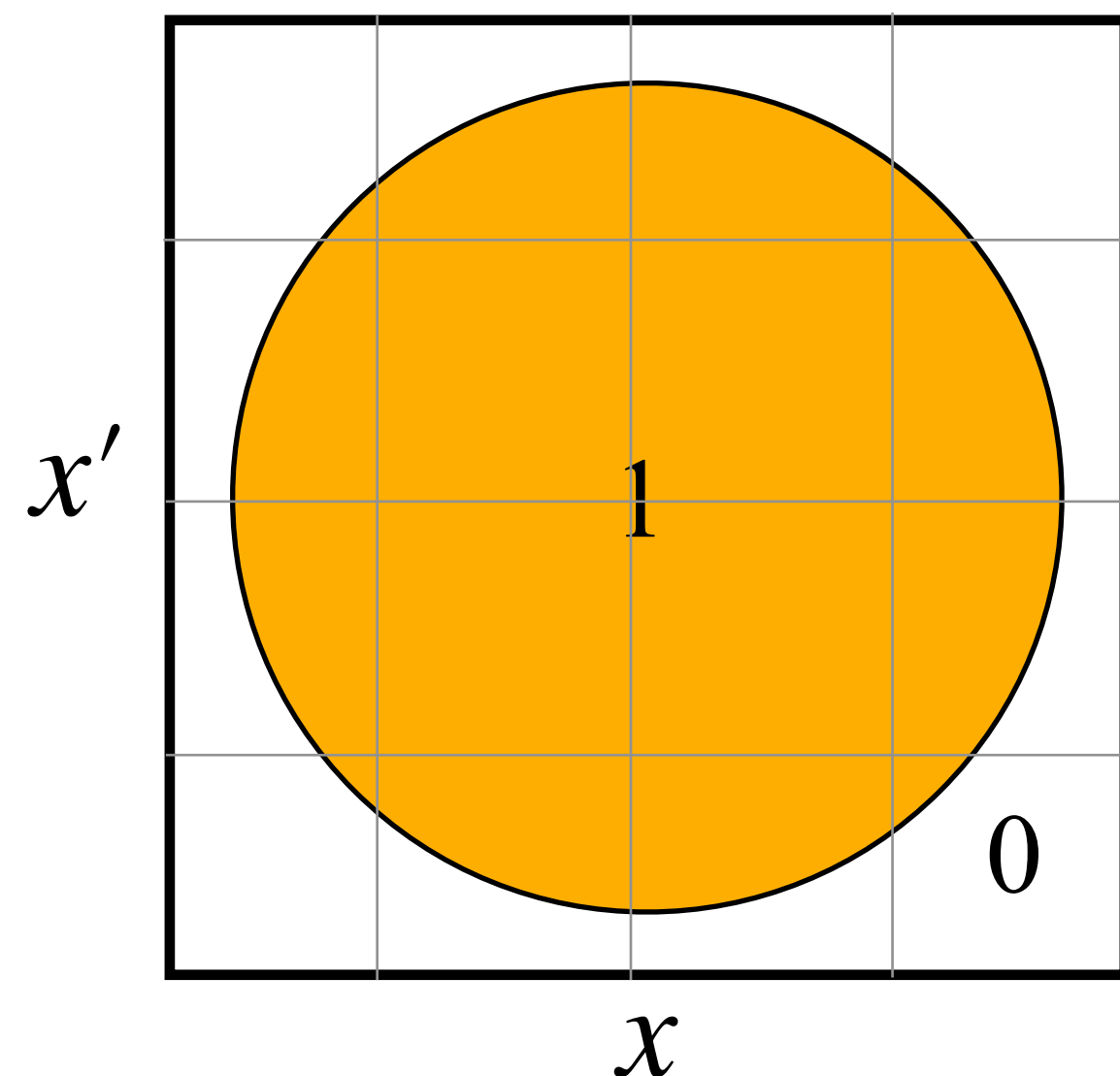


Representation of Continuous Functions: <https://tensornetwork.org/functions/>

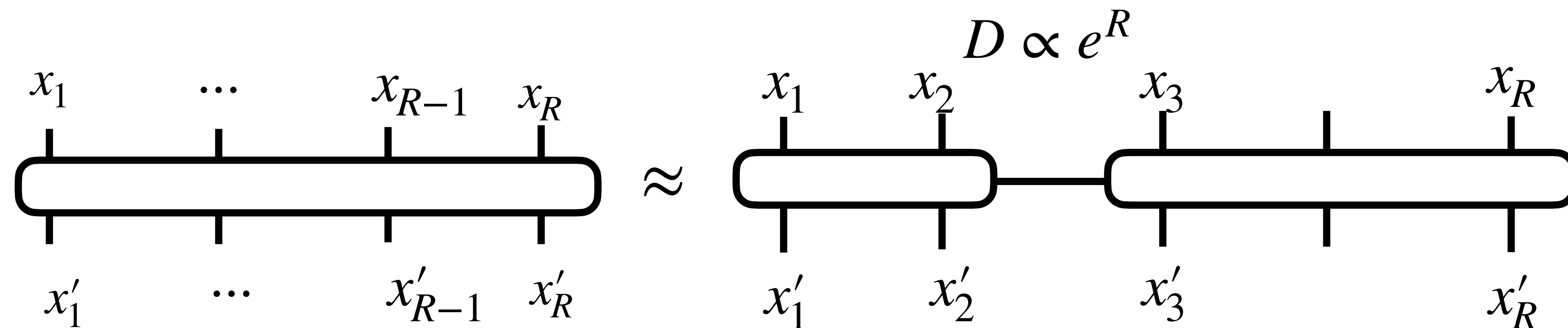
Written by Miles Stoudenmire

PRX 13, 021015 (2023), arXiv:2303.11819

Non-compressible multivariate functions



Many linearly independent patches



Singular value decomposition

My naive understanding

Step functions with gradually changing angles are not compressible.

Broadening and/or the fixed number of angles may help.

Line discontinuity → OK

or, in other words, if A is partitioned into $2^k \times 2^k$ blocks B_{pq} , $p, q = 1, \dots, 2^{d-k}$, then r_k is equal to the dimension of linear space spanned by matrices B_{pq} .

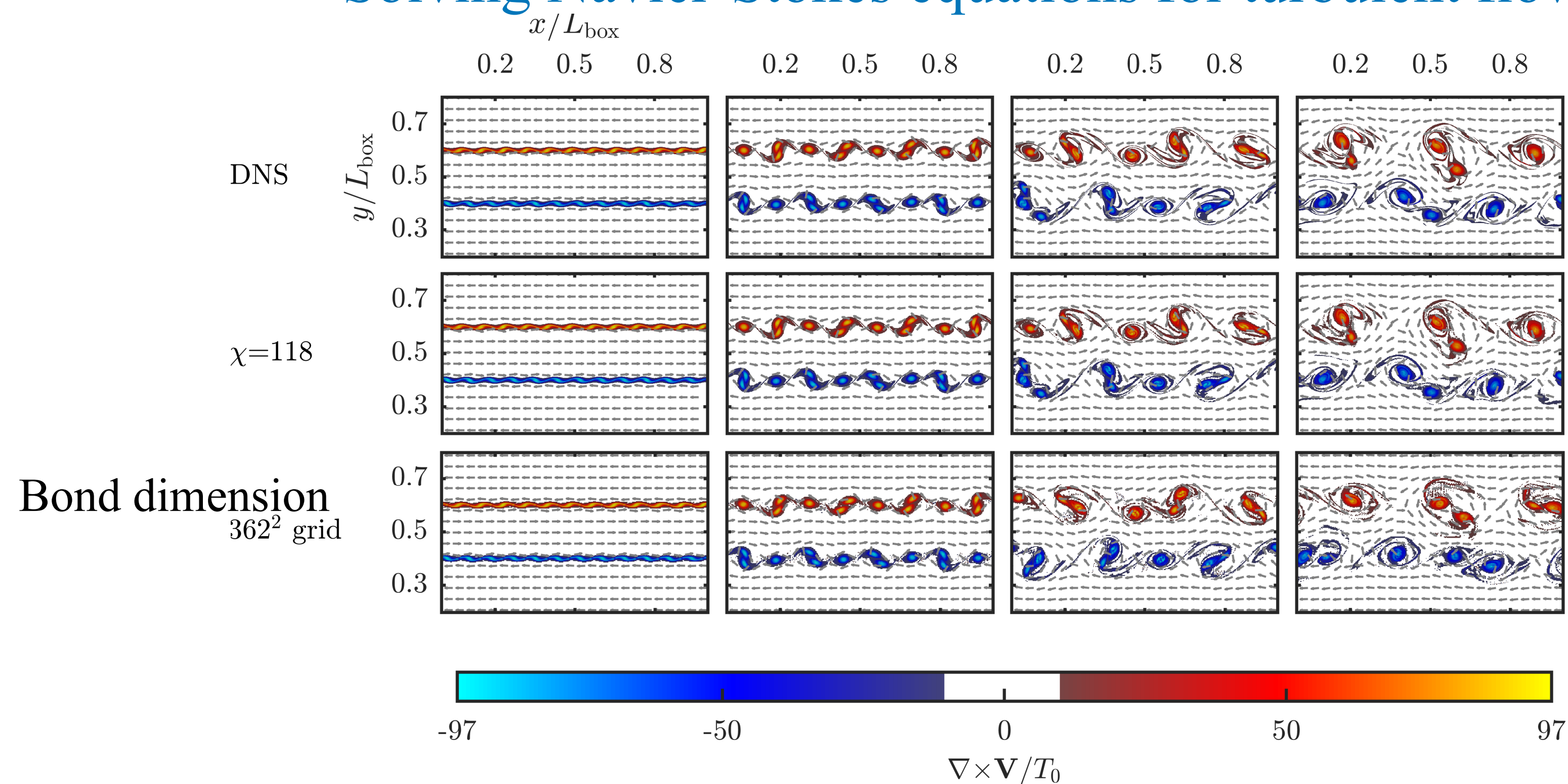
For this example, it is easy to see that each splitting into $2^k \times 2^k$ blocks produces $\mathcal{O}(n)$ different linearly independent blocks. Therefore

$$r_k = \mathcal{O}(n),$$

and the memory to store matrices U_1, \dots, U_d in this case is $\mathcal{O}(n^2)$; thus no compression is provided for this simple example. However, the WTT transform with limited maximal ranks works well. For this example, there is a natural accuracy level related to discrete representation of a circle on a rectangular grid, roughly $\frac{1}{n}$, and that is confirmed by experiment. If the threshold parameter ε is set to 10^{-2} , the number of

Recent applications of QTT in physics

Solving Navier-Stokes equations for turbulent flows



N. Gourianov *et al.*, Nat. Comput. Sci. **2**, 30 (2022)

Vlasov-Poisson equations for collisionless plasmas

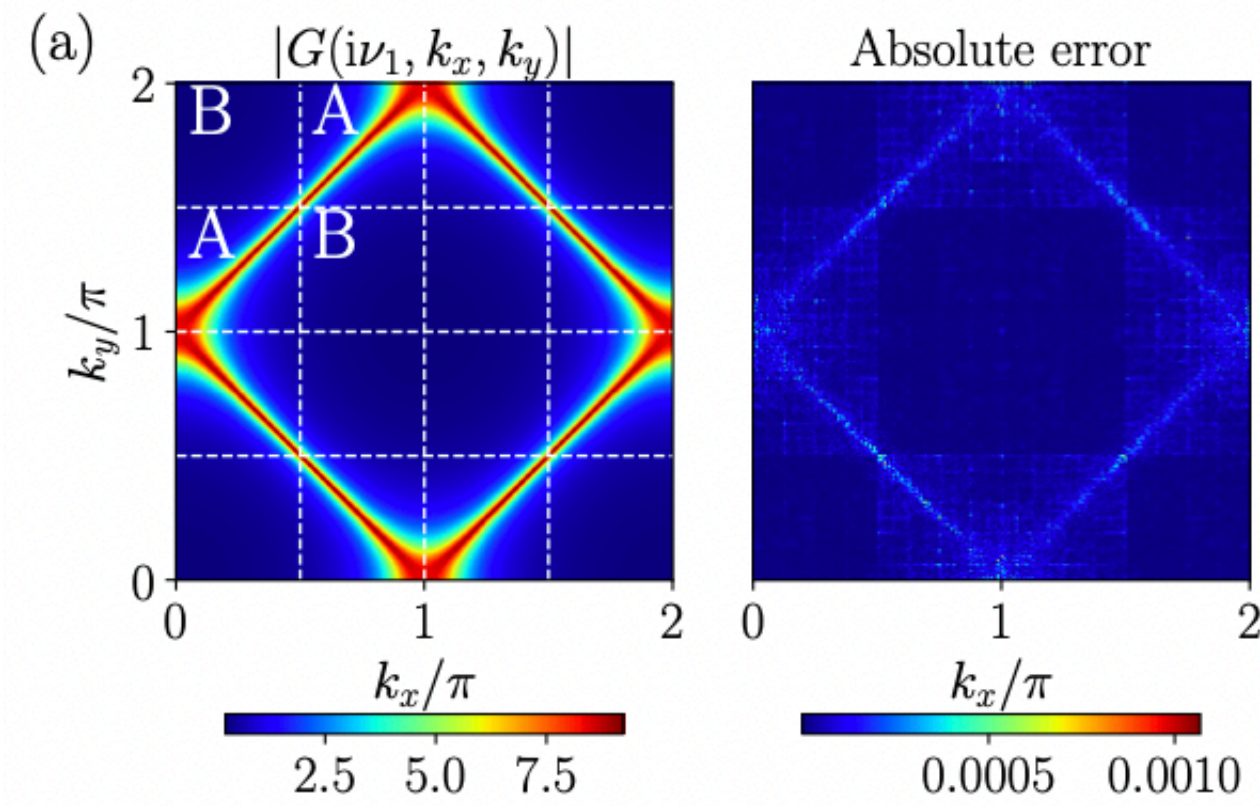
E. Ye and N. F. G. Loureiro, arXiv:2205.11990

This study: Are correlation functions of **quantum** systems compressible?

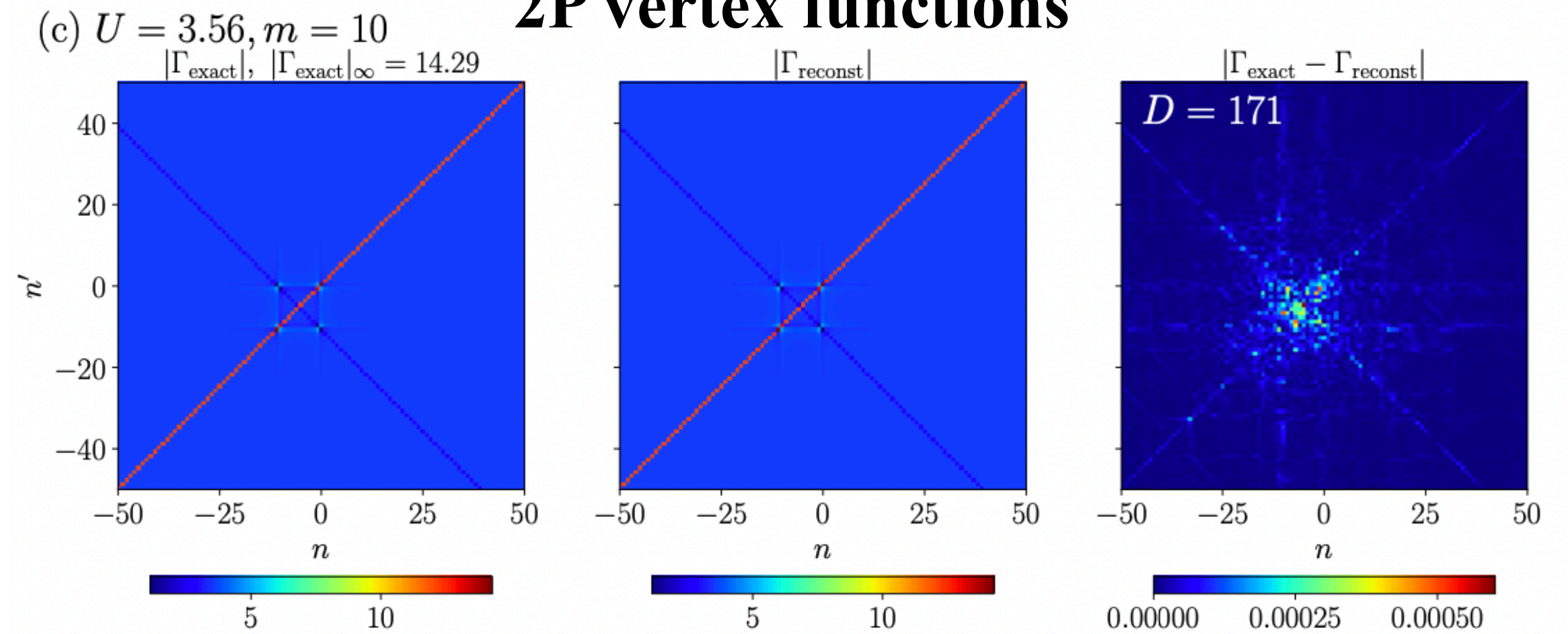
Compression

Compression

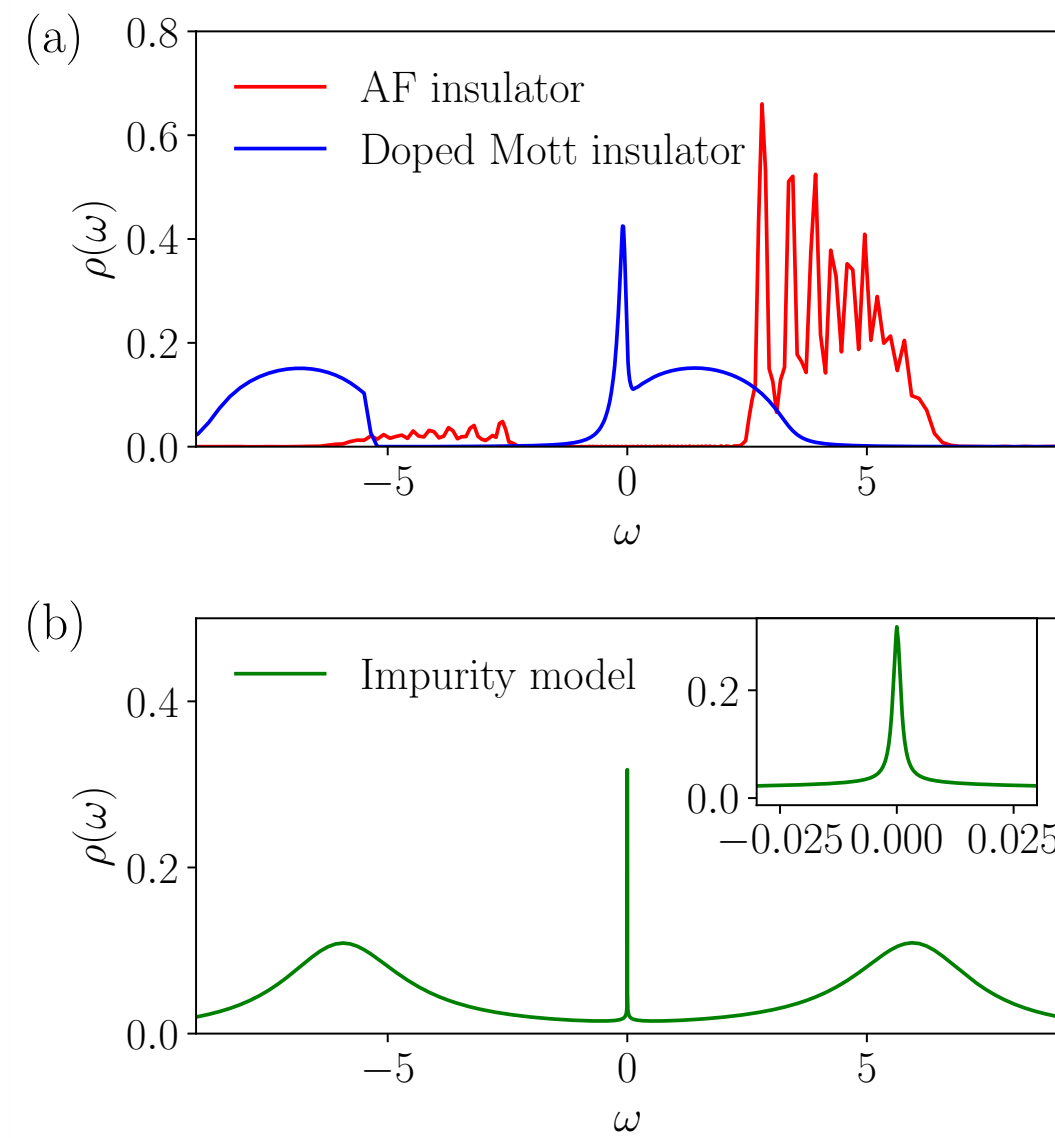
1P momentum space



2P vertex functions



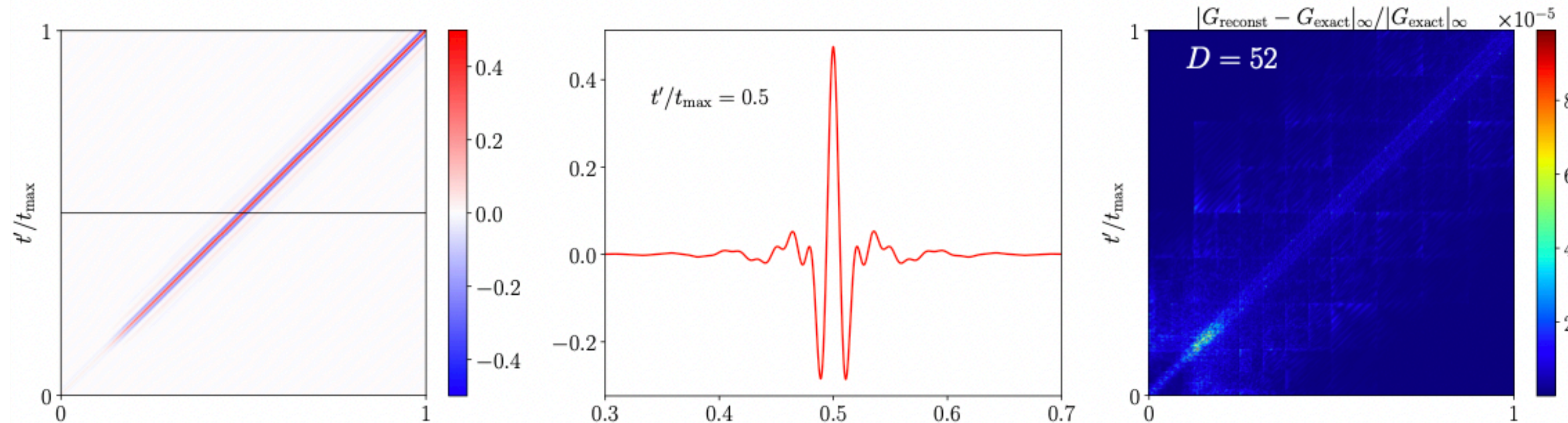
Spectral function



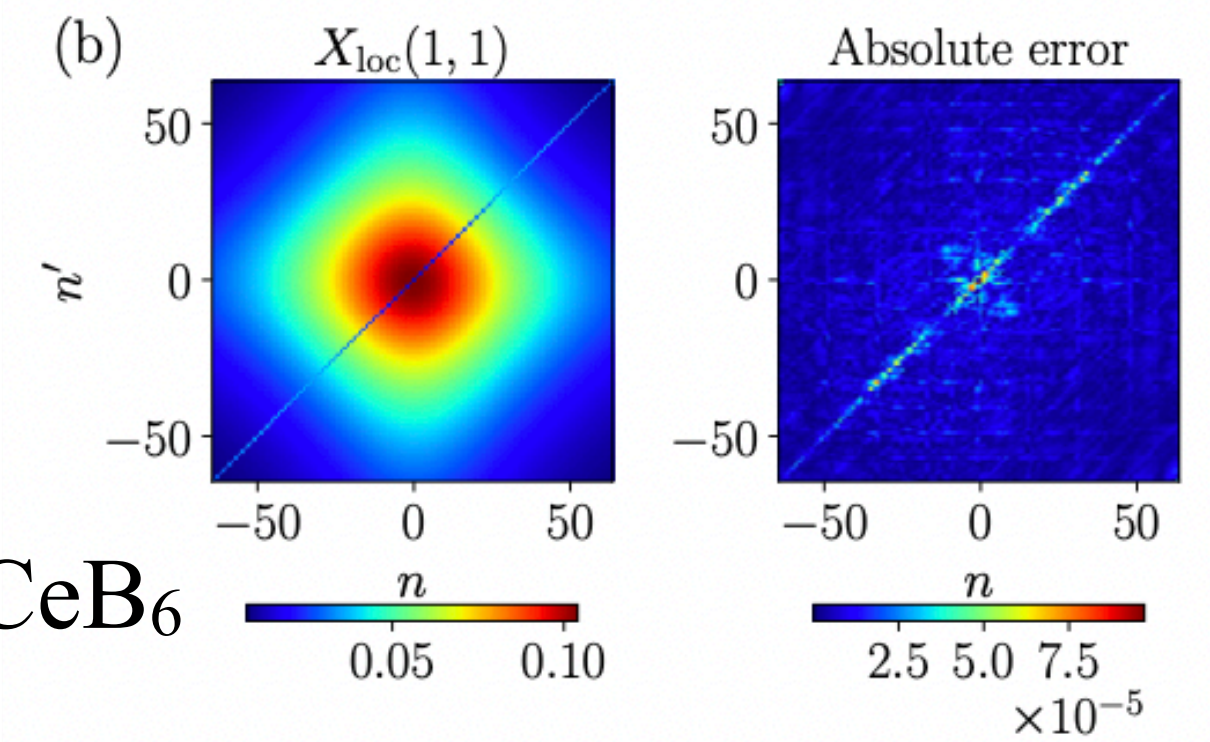
All QTT compressible!

HS *et al.*, PRX 13, 021015 (2023)

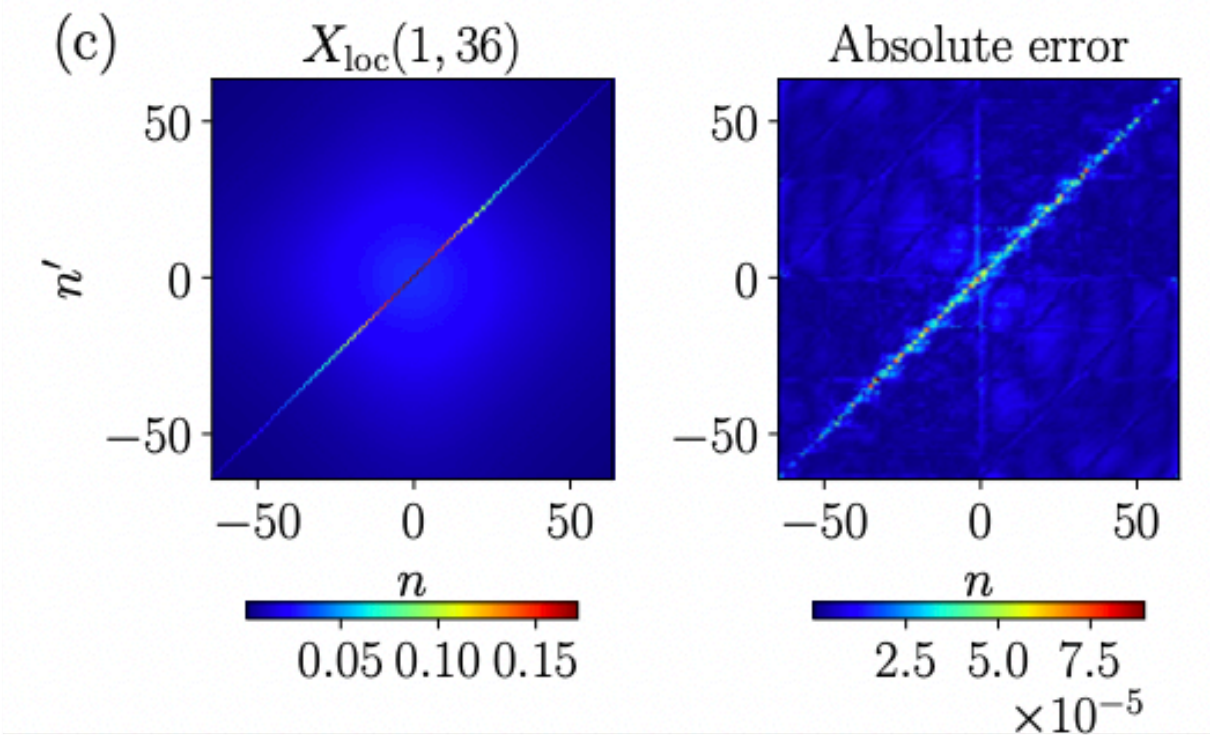
Nonequilibrium system (real-time Green's function)



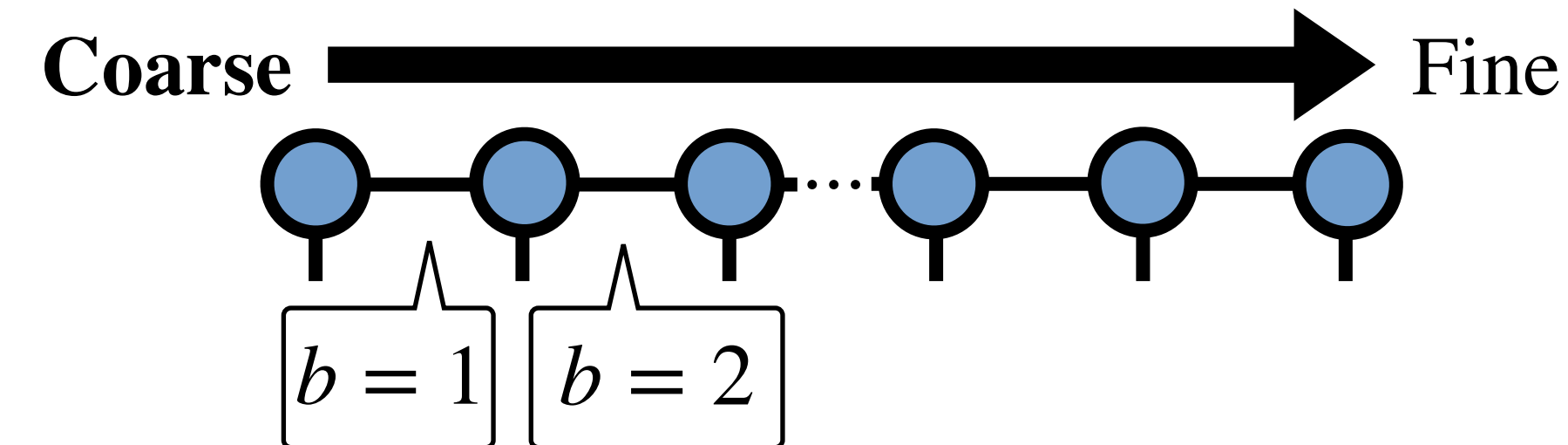
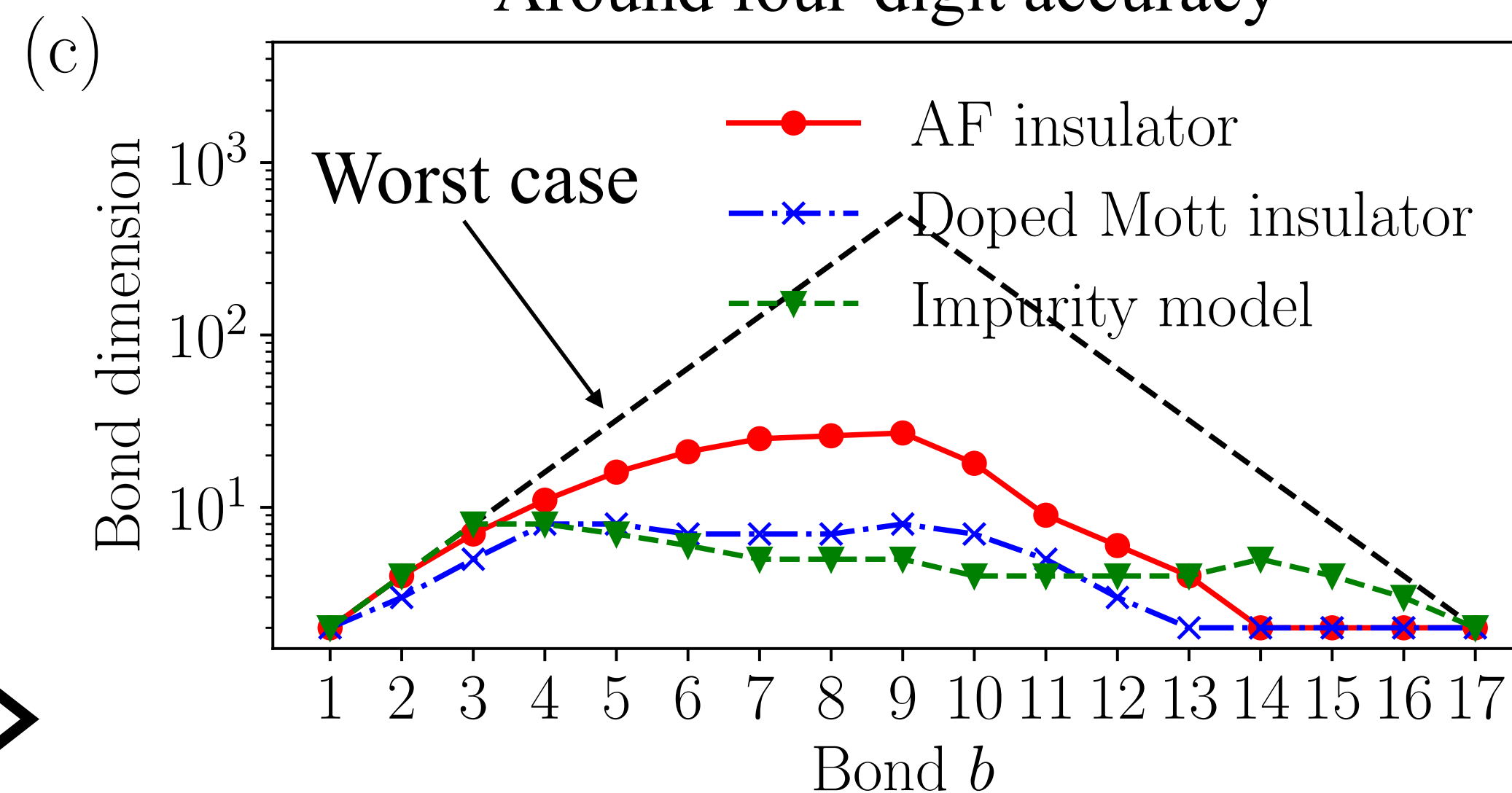
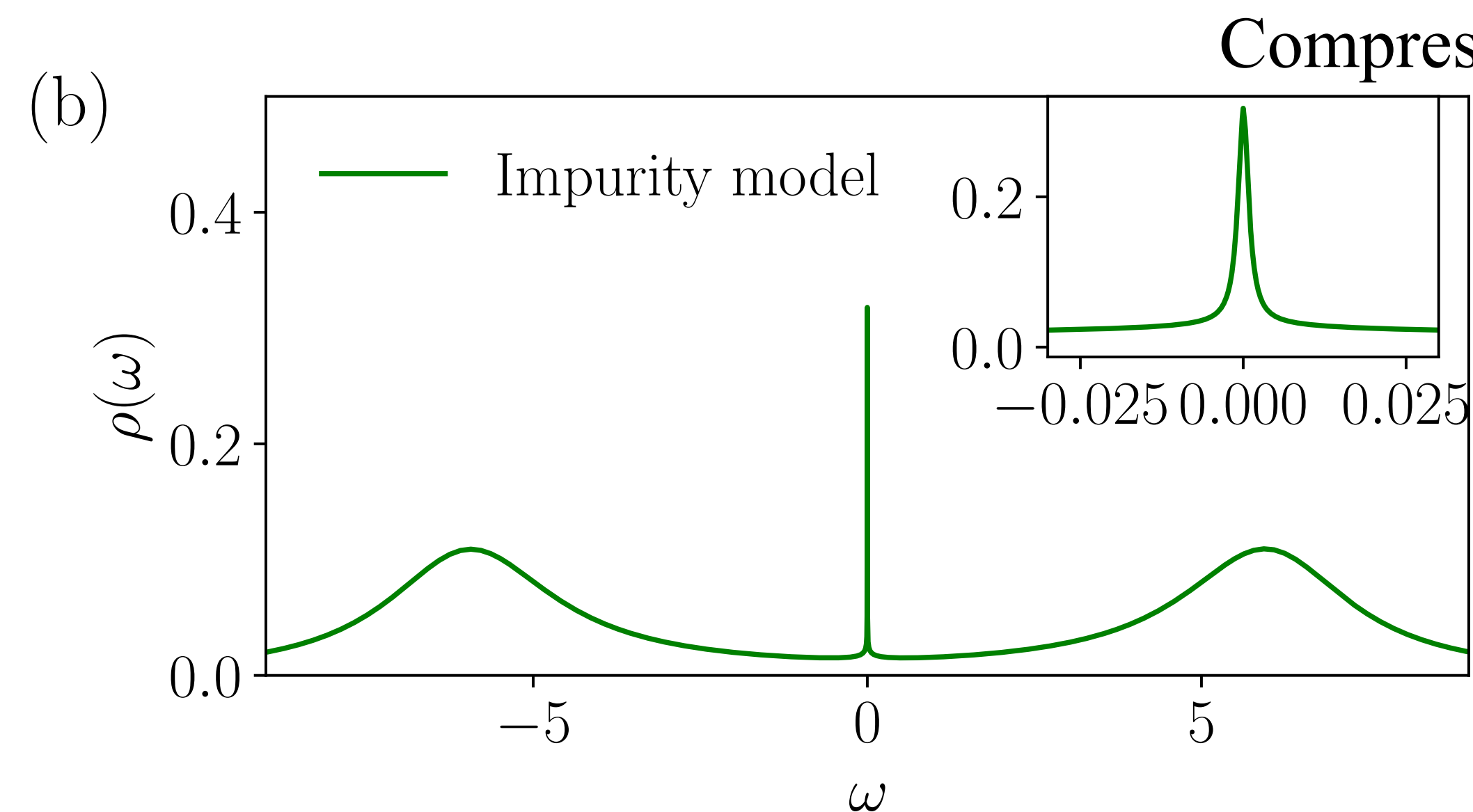
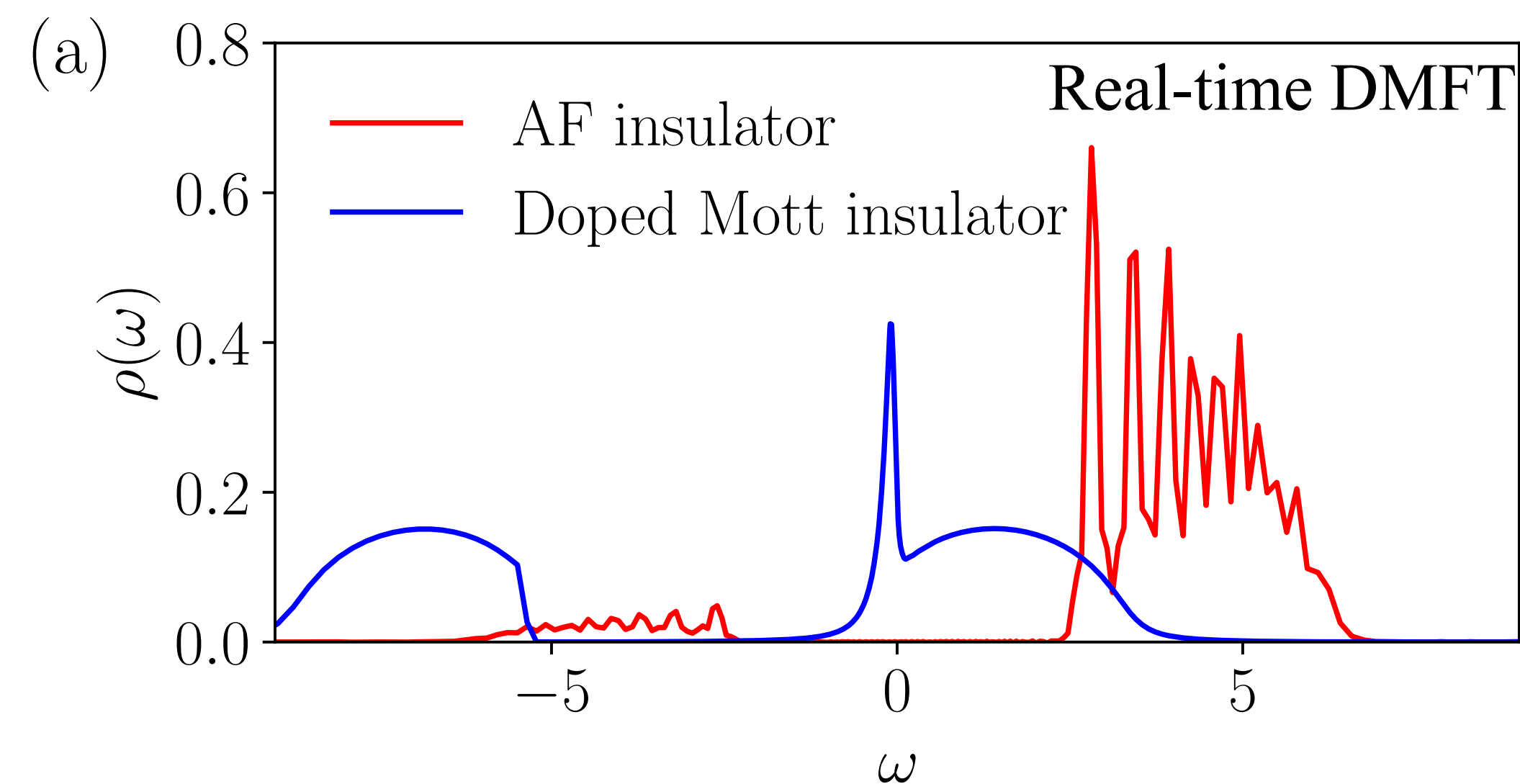
Multipolar susceptibility



CeB₆



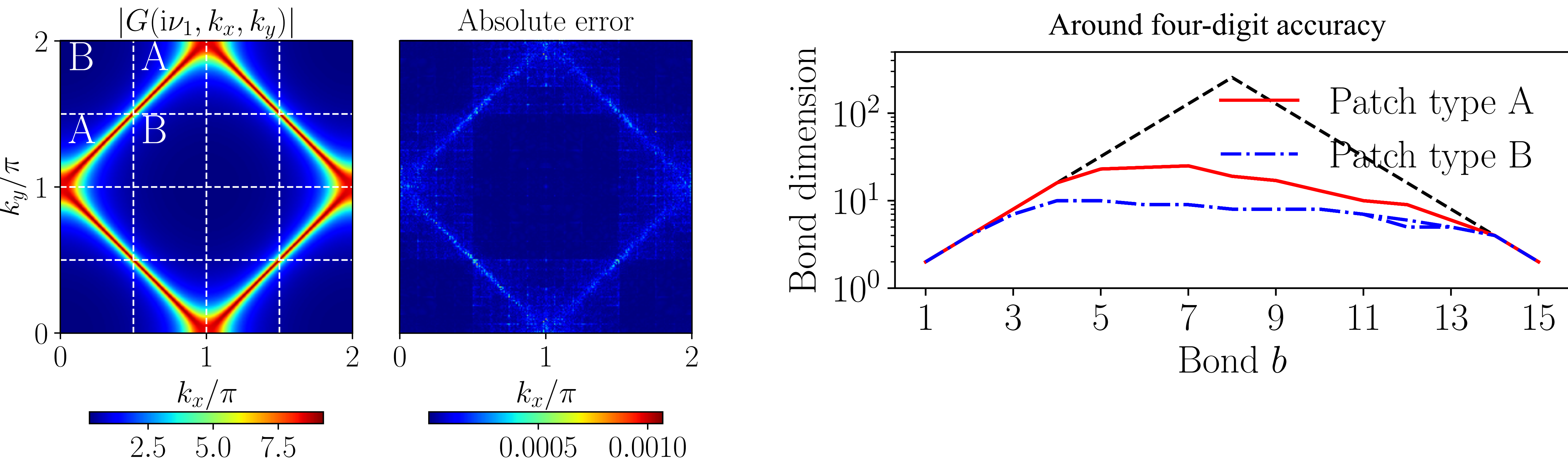
Spectral functions



- Sharp peaks can be represented.
- Larger bond dimension for more features

Momentum-resolved Green's function

Hubbard model, $T = 0.03$, $U = 1.1$ (band width: 8), FLEX approximation



Our conjecture: $D \propto \beta^{\frac{d-1}{2}}$ (d : spatial dimension)

$\beta = 1/T$

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, [HS](#), X. Waintal, arXiv:2303.11819

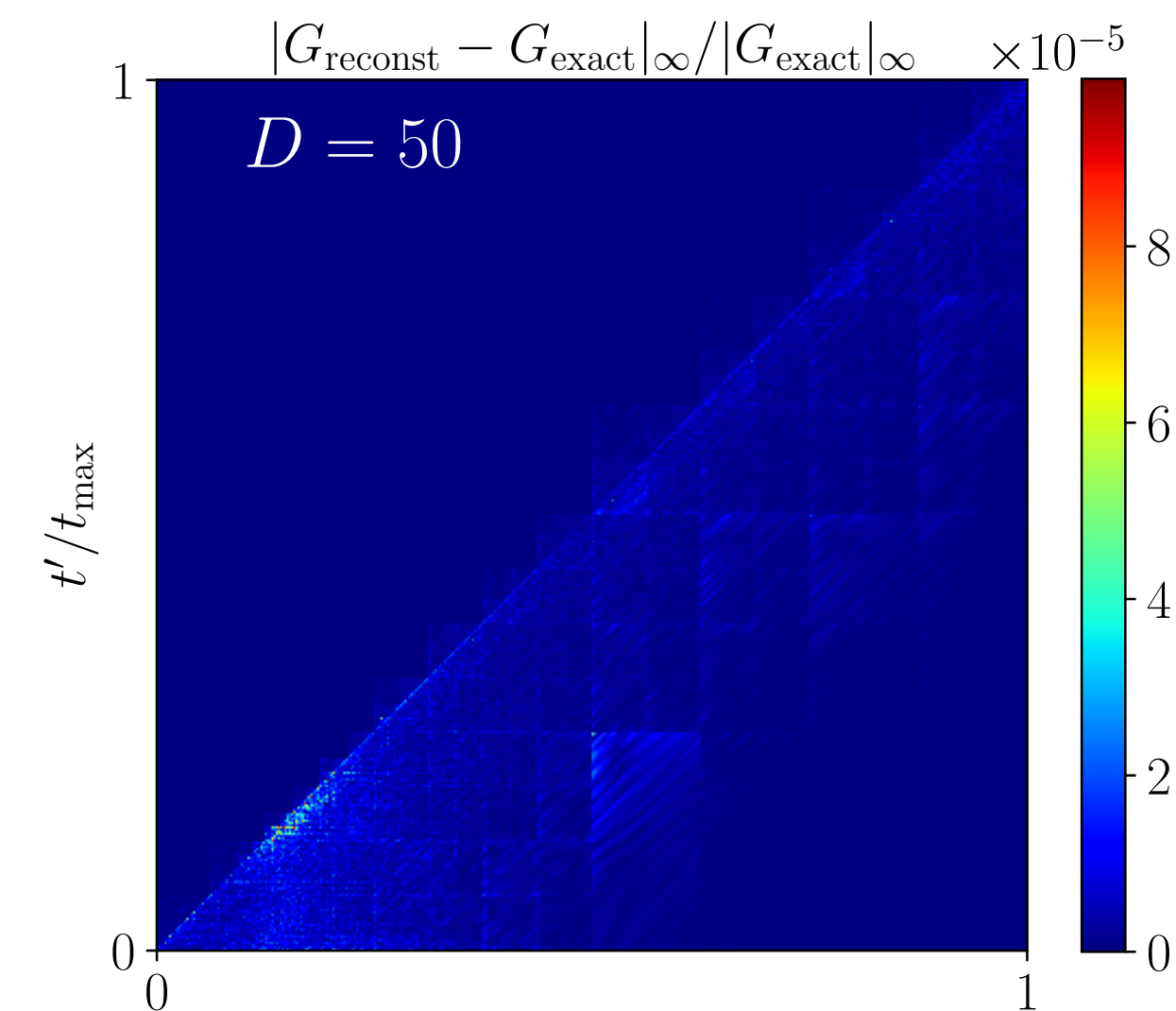
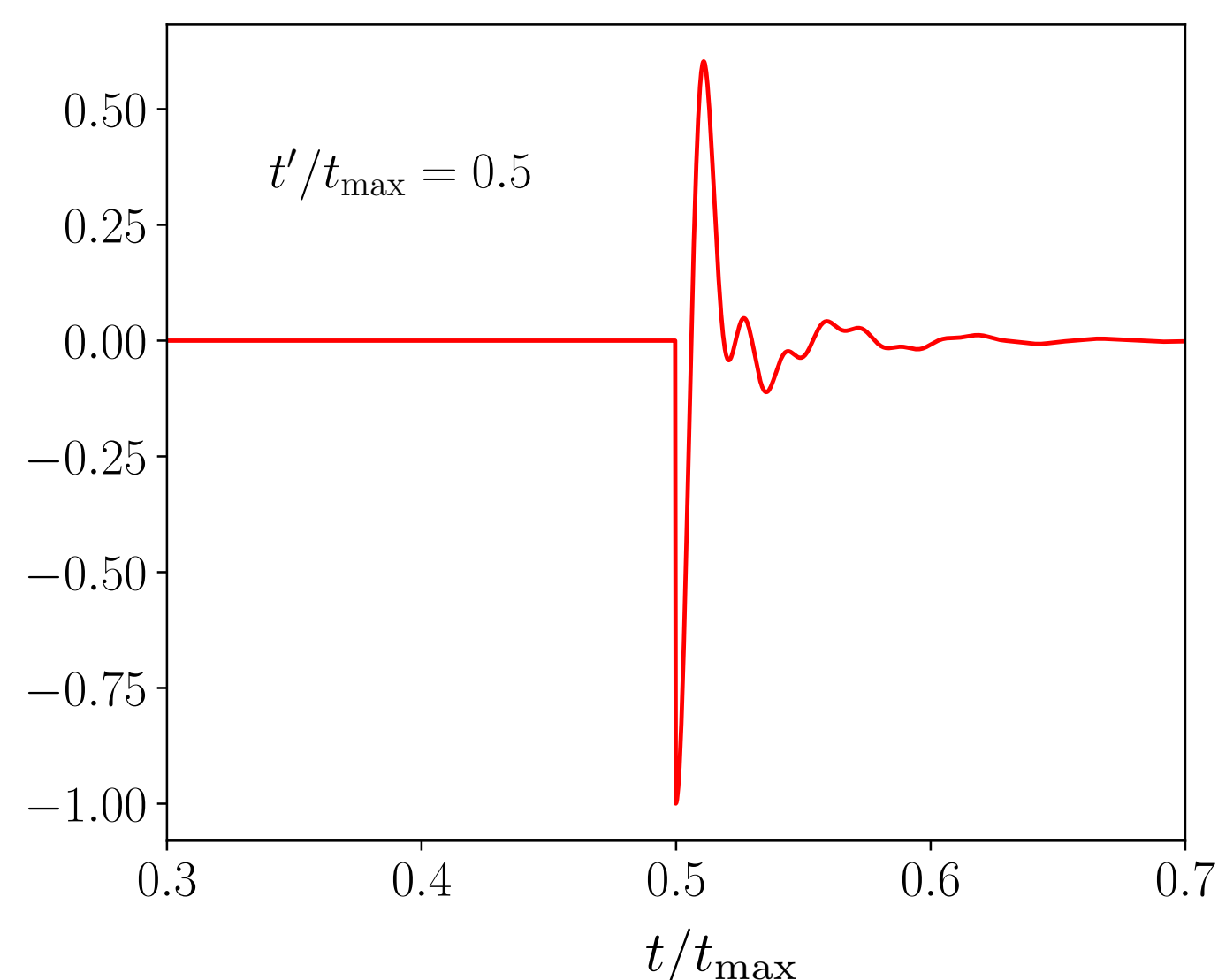
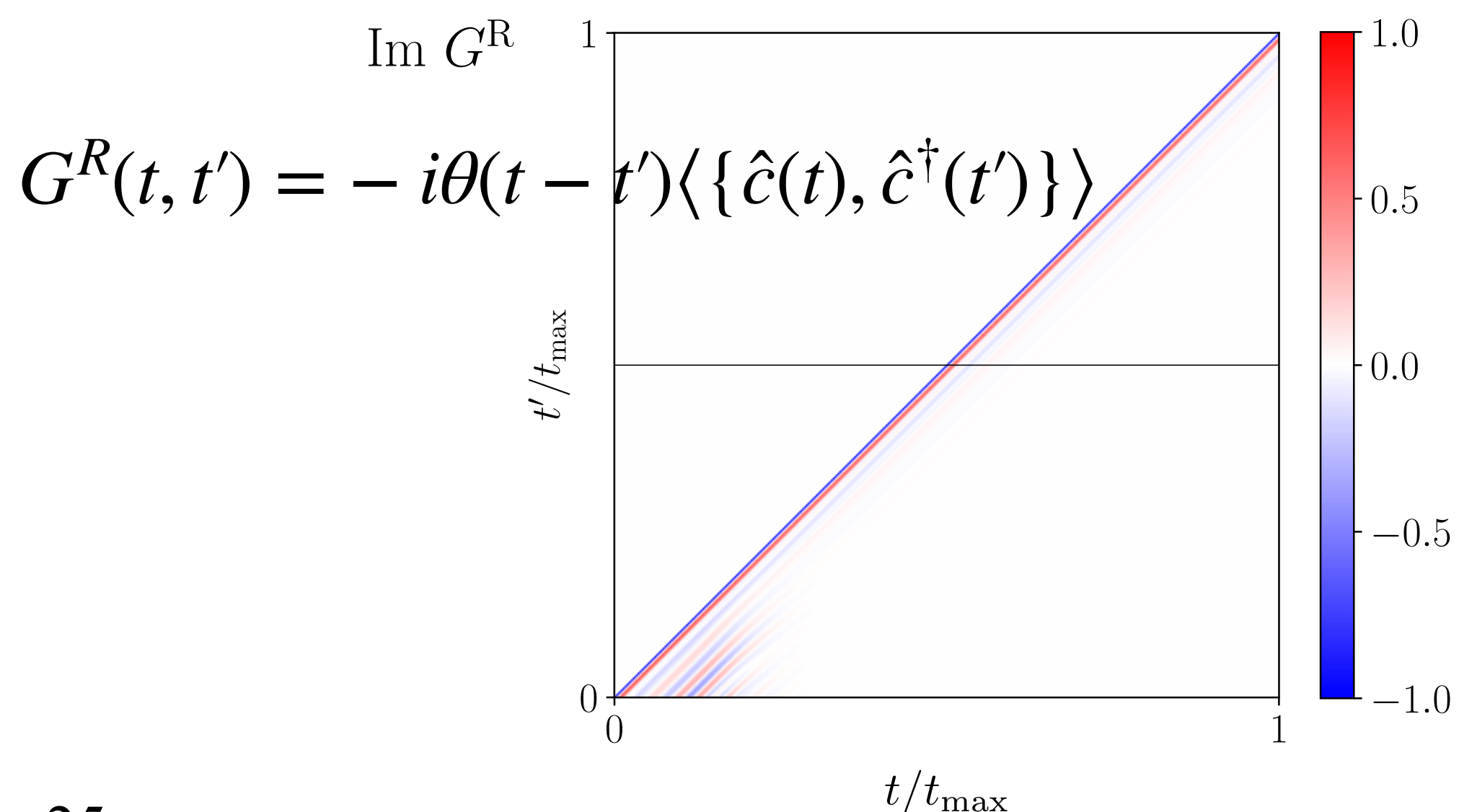
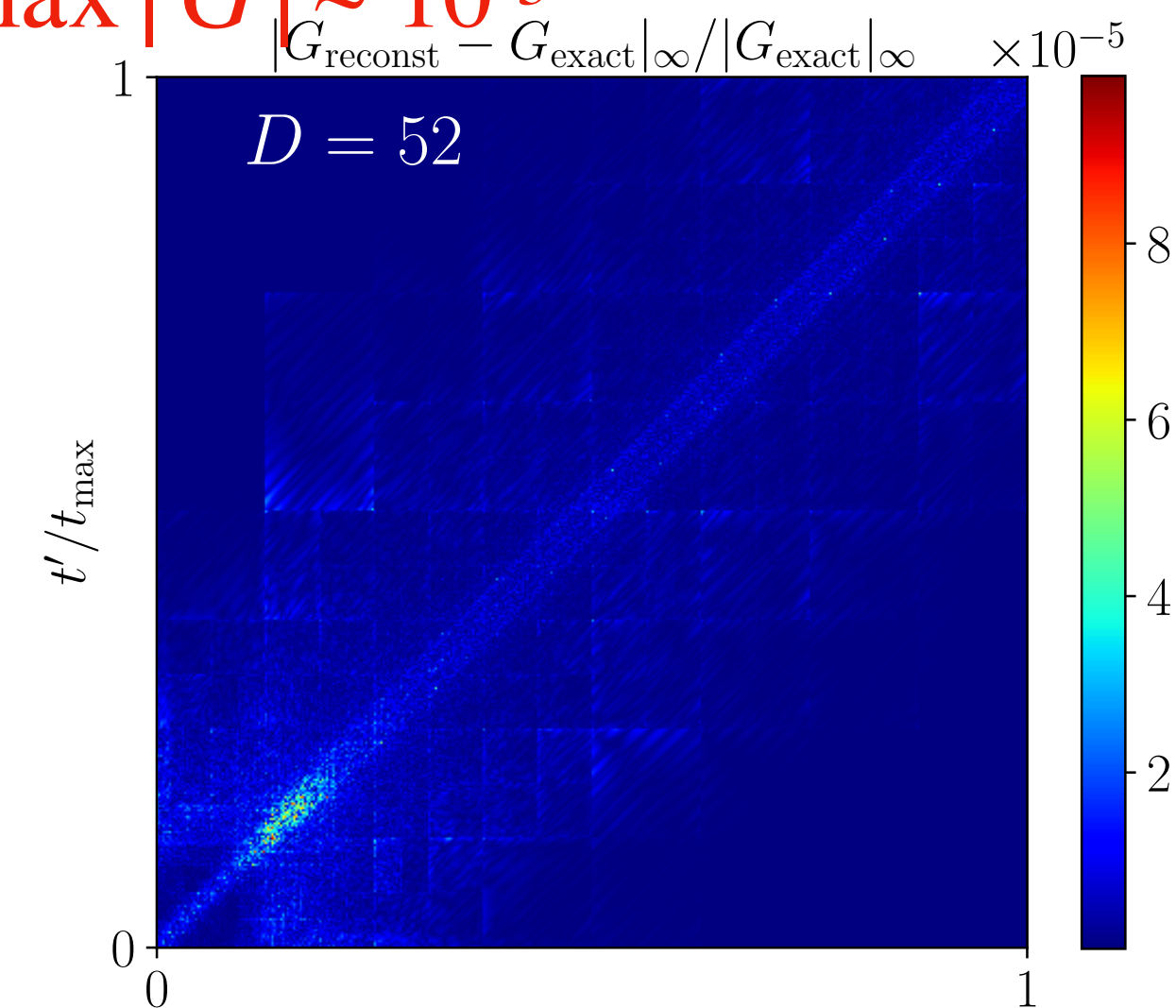
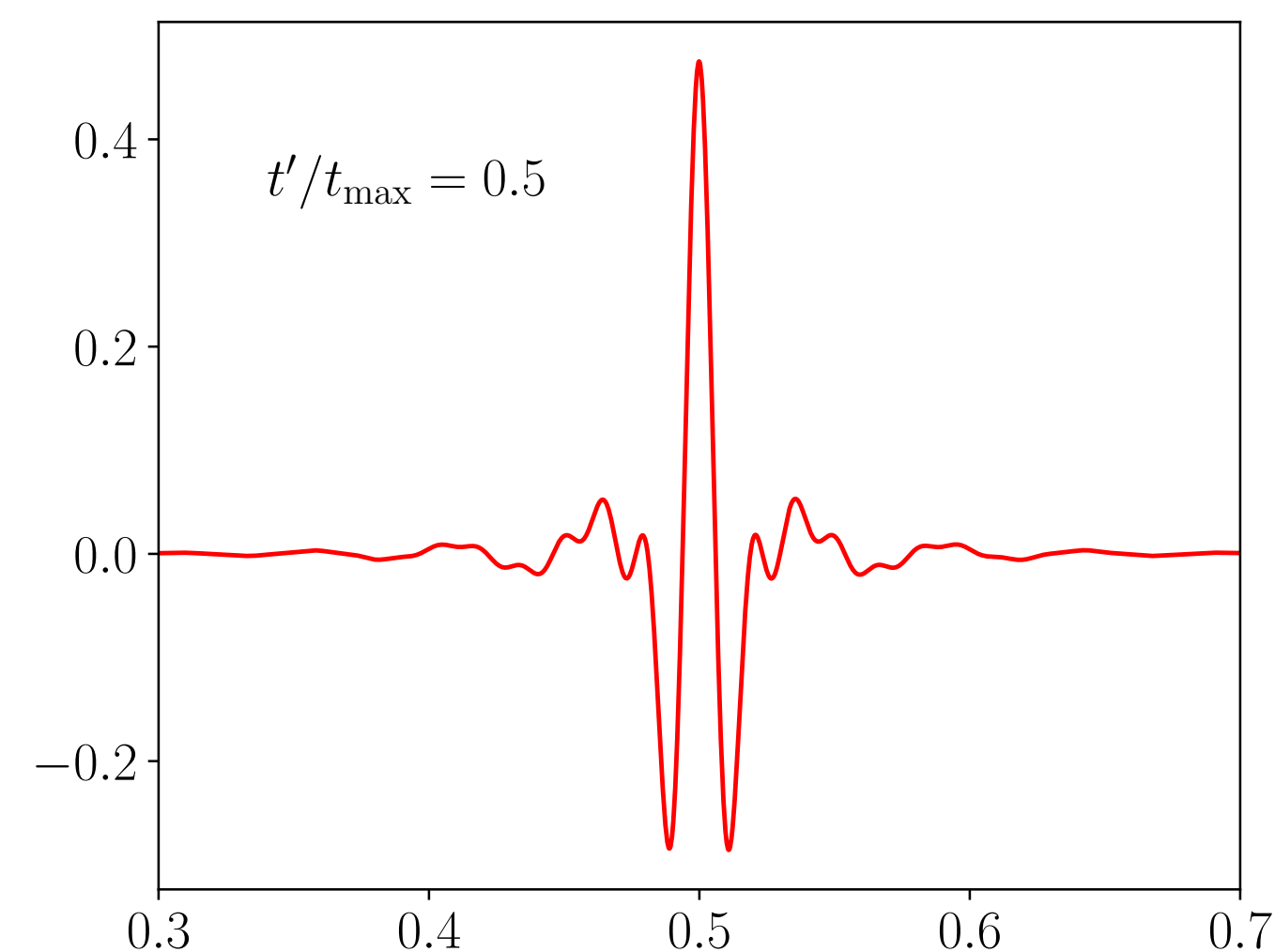
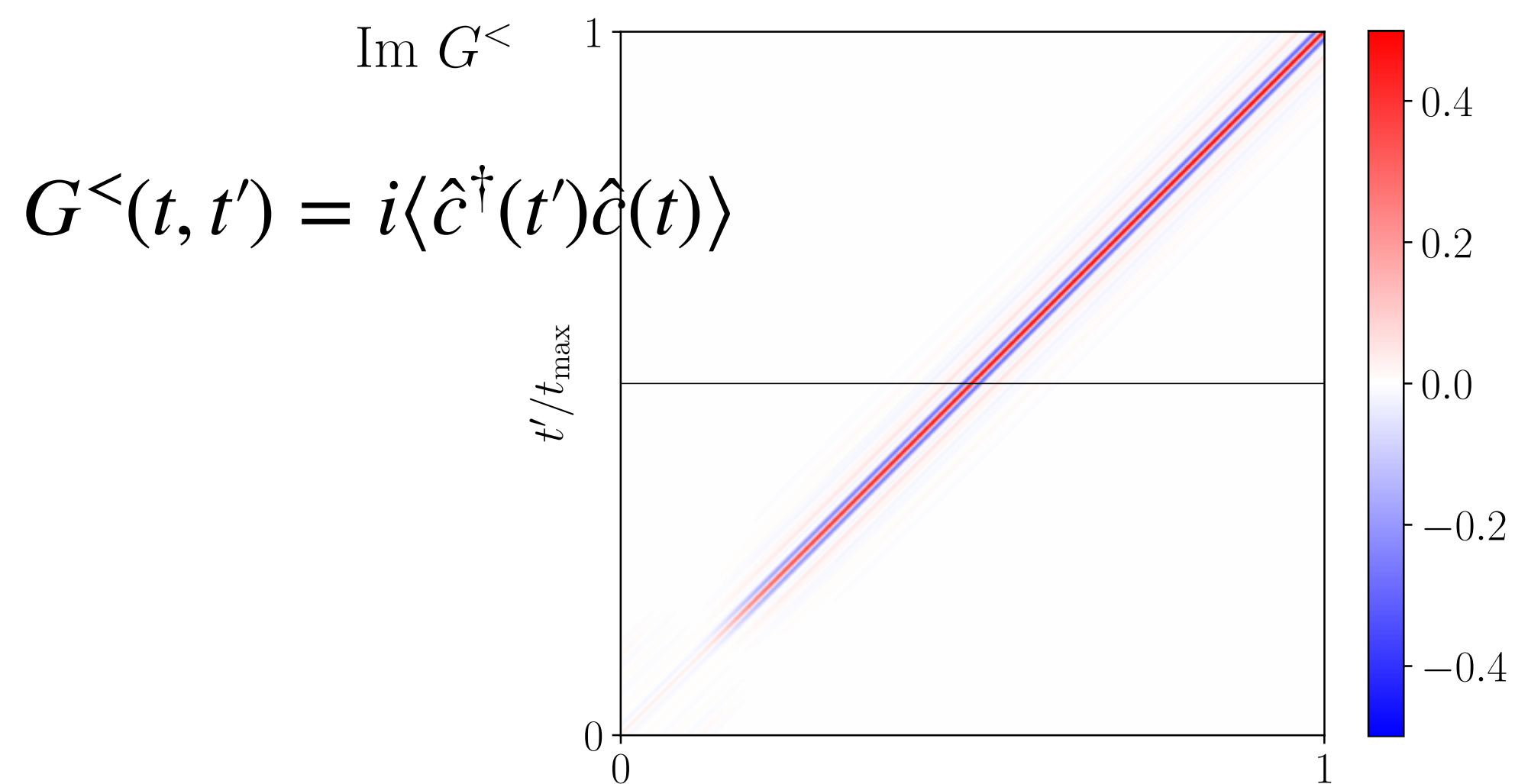
Real-time Green's function: Non-equilibrium case

Low- T AF Mott phase excited by a short electric field pulse, Bethe lattice, $T = 0.05$

Compression ratio $\sim 10^3$

$|\delta G|/\max |G| \sim 10^{-3}$

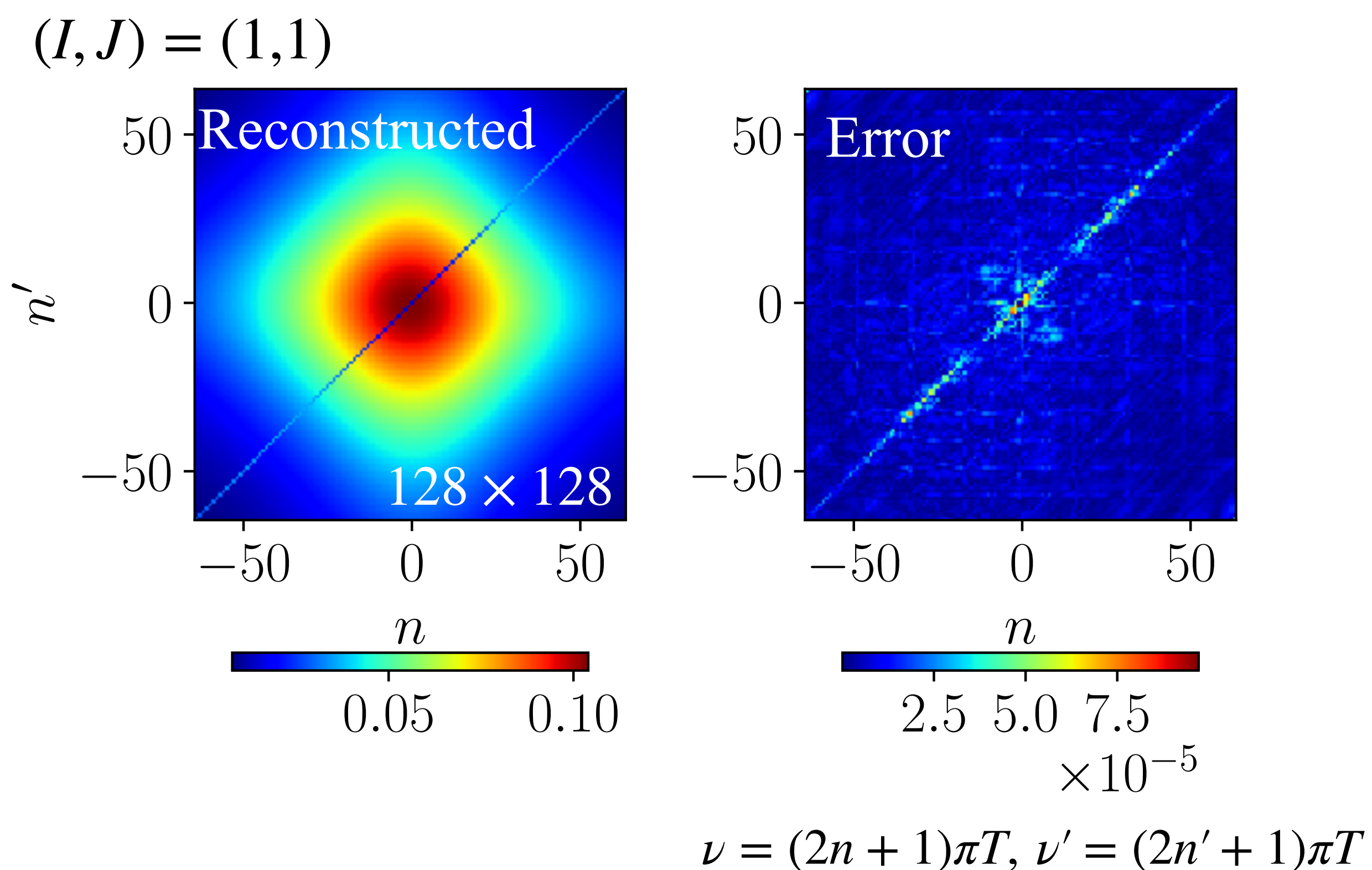
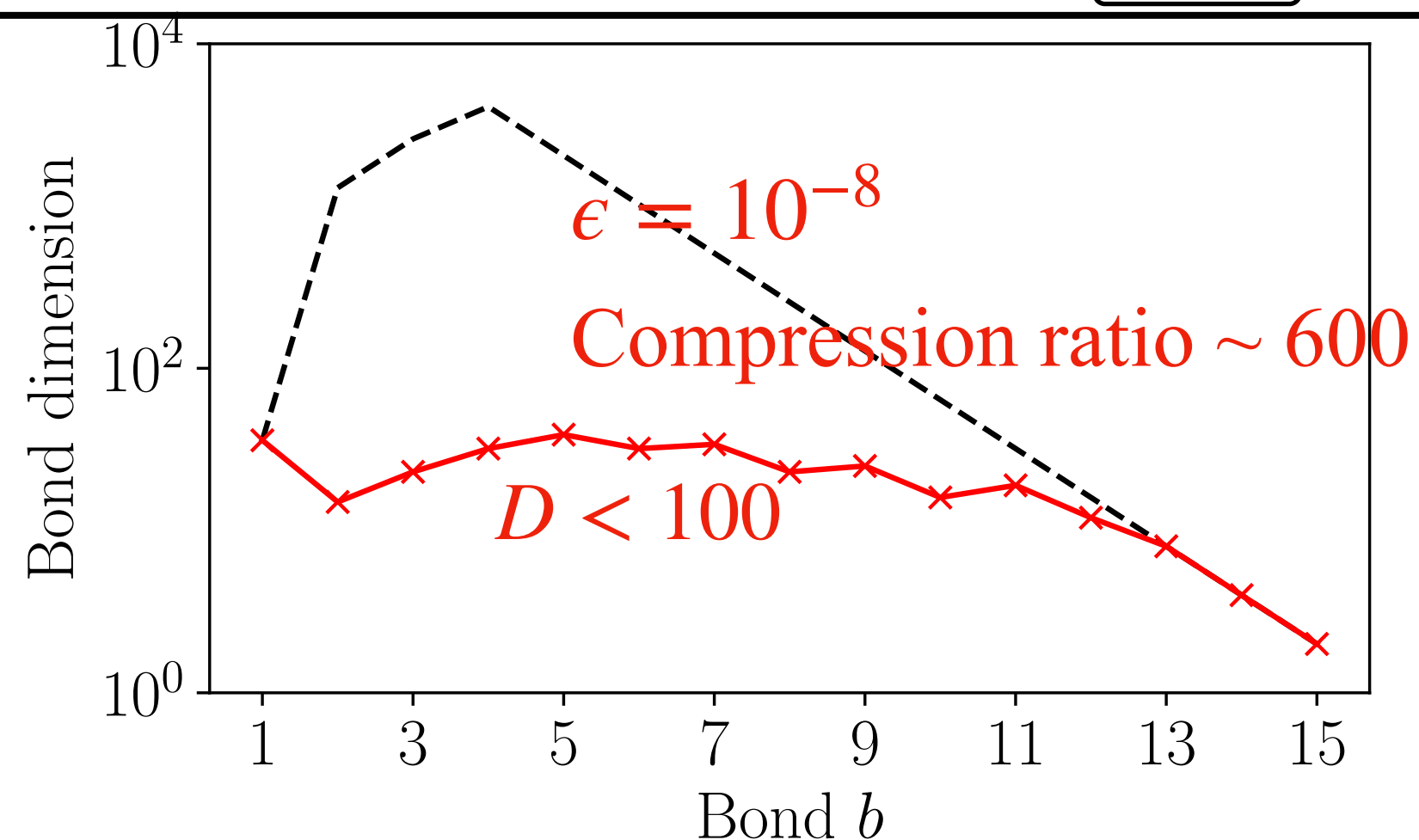
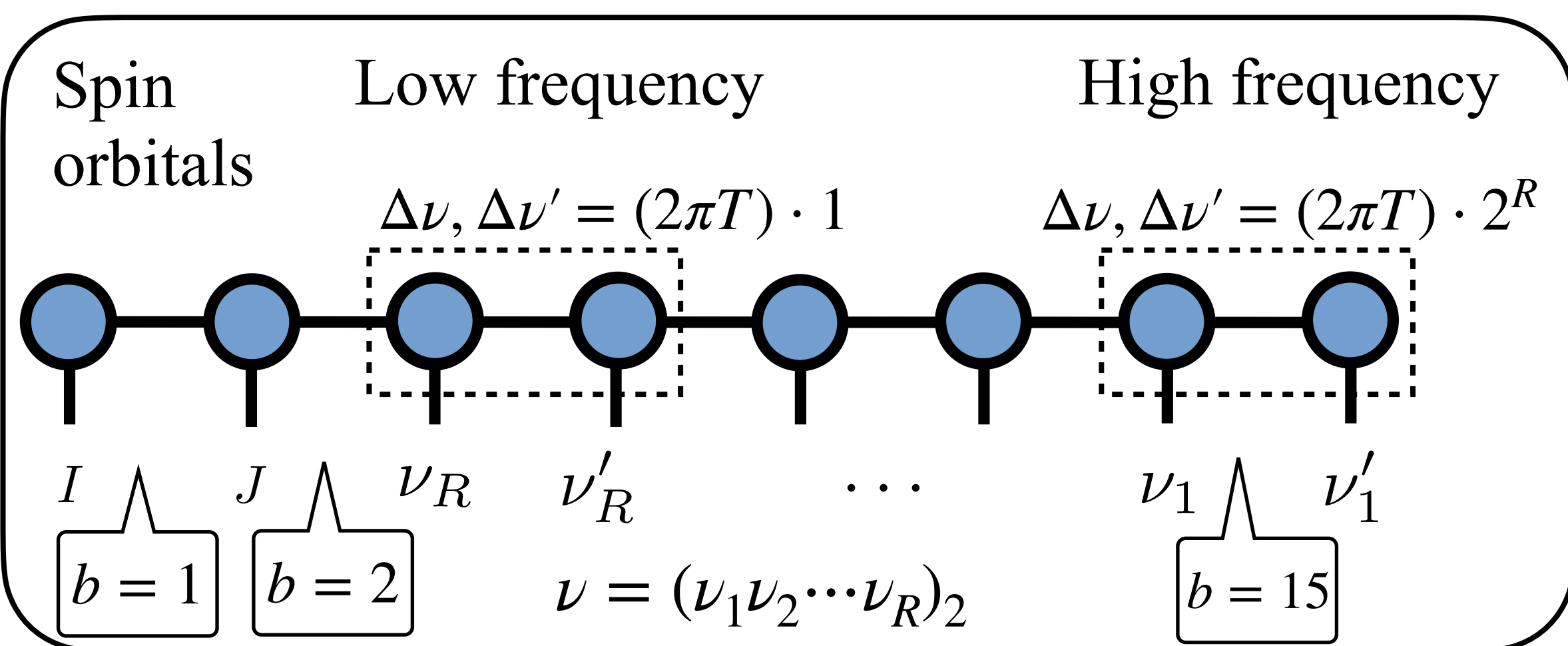
$|G_{\text{reconst}} - G_{\text{exact}}|_{\infty}/|G_{\text{exact}}|_{\infty} \times 10^{-5}$



Multipolar susceptibility of an f -electron system: CeB_6

J. Otsuki, K. Yoshimi, **HS**, and H. O. Jeschke, arXiv:2209.10429v1

- Six correlated states ($j=5/2$)
- DFT+DMFT using the Hubbard-I approximation
- Static multipolar susceptibility computed by solving Bethe-Salpeter equation



Multipolar susceptibility

J. Otsuki, K. Yoshimi, **HS**, and H. O. Jeschke, arXiv:2209.10429v1

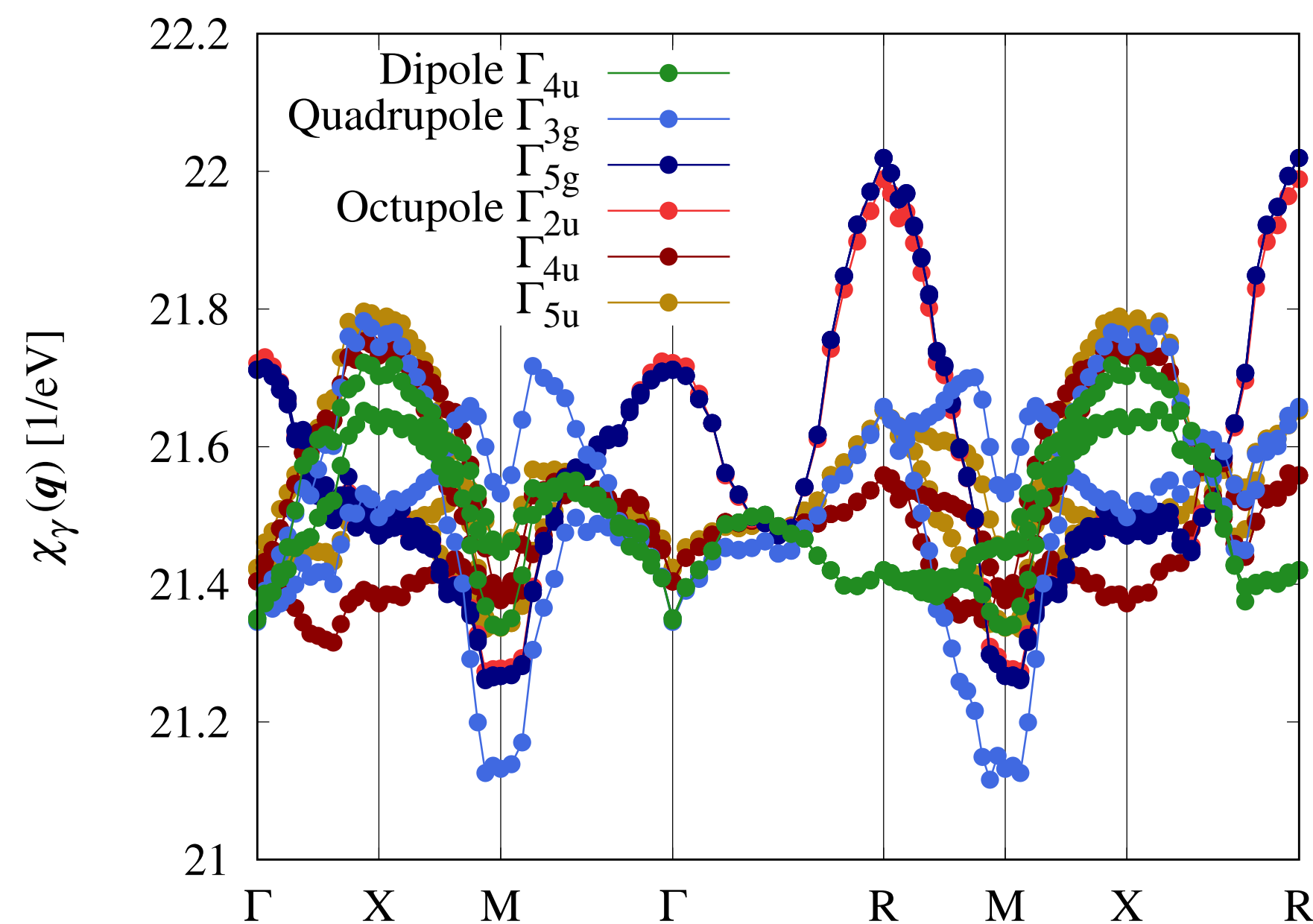
Coarse \longrightarrow Fine



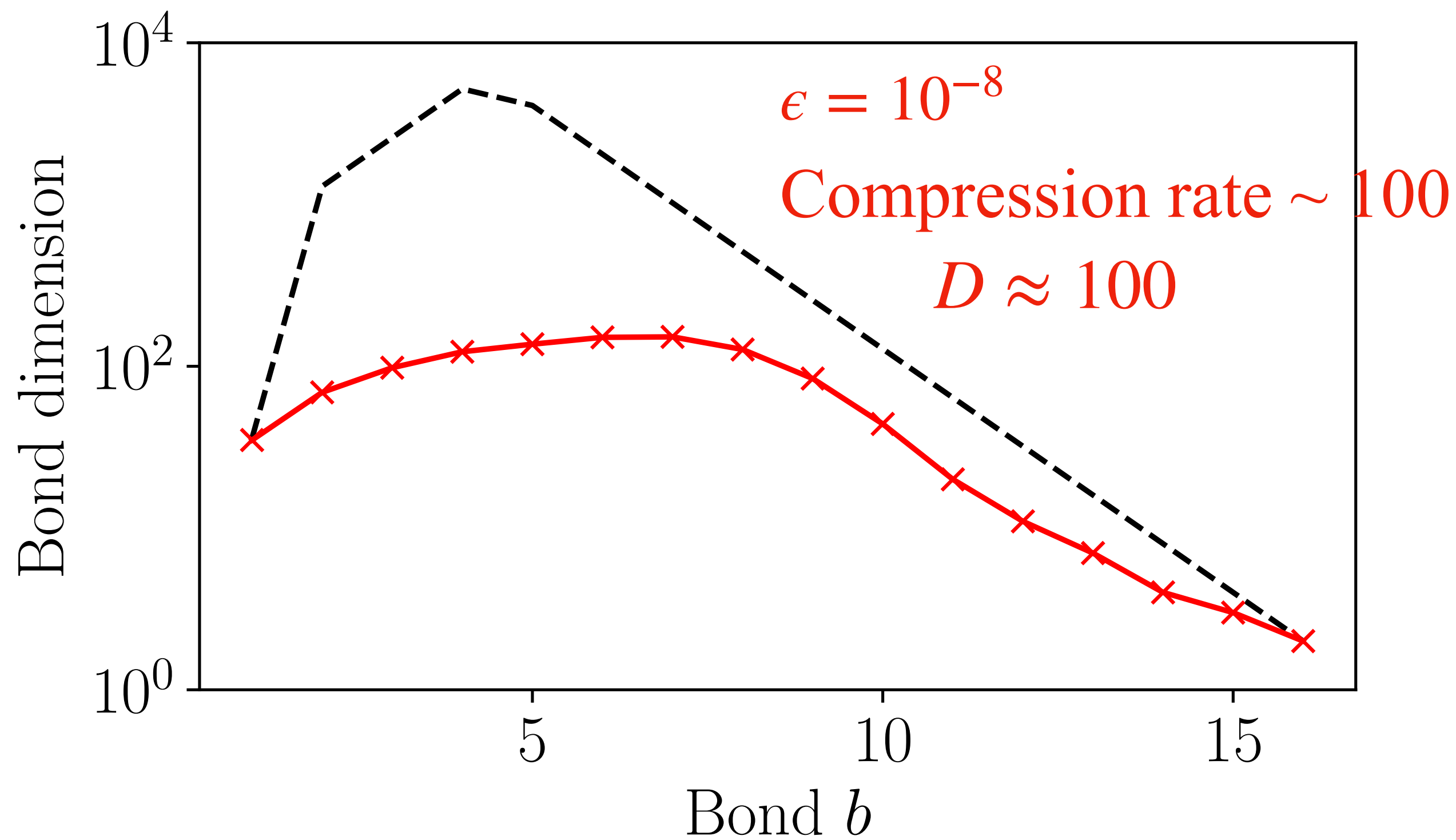
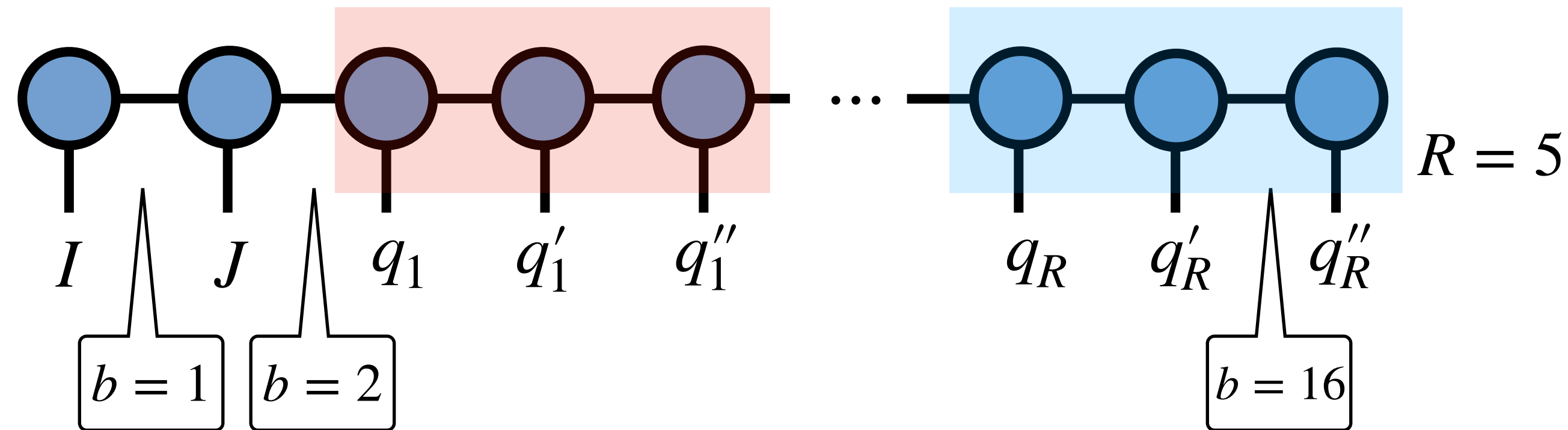
I J q q' q''

36 36 32 32 32

Spin orbital Momentum



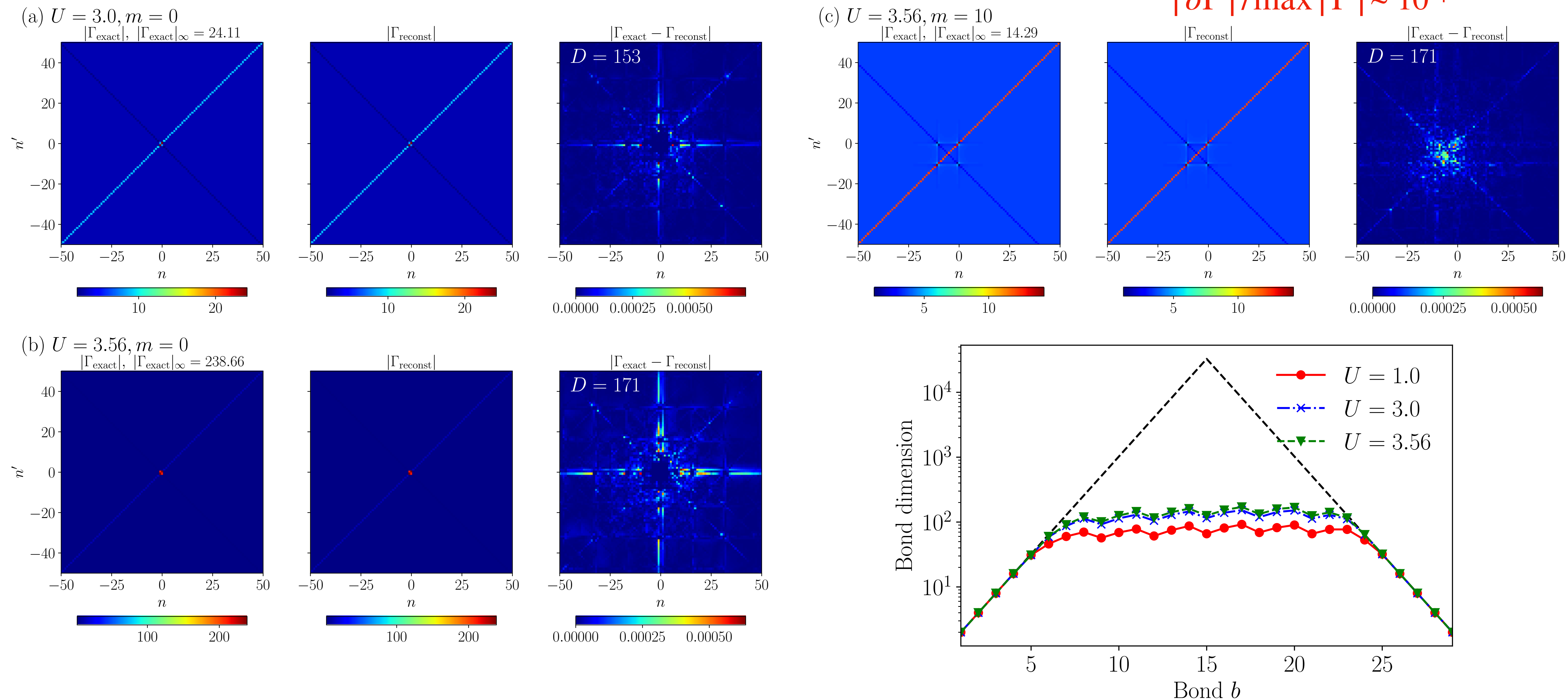
J. Otsuki, K. Yoshimi, **HS**, and H. O. Jeschke, arXiv:2209.10429v1



PRX **13**, 021015 (2023), arXiv:2303.11819

Three-frequency vertex function: Hubbard atom

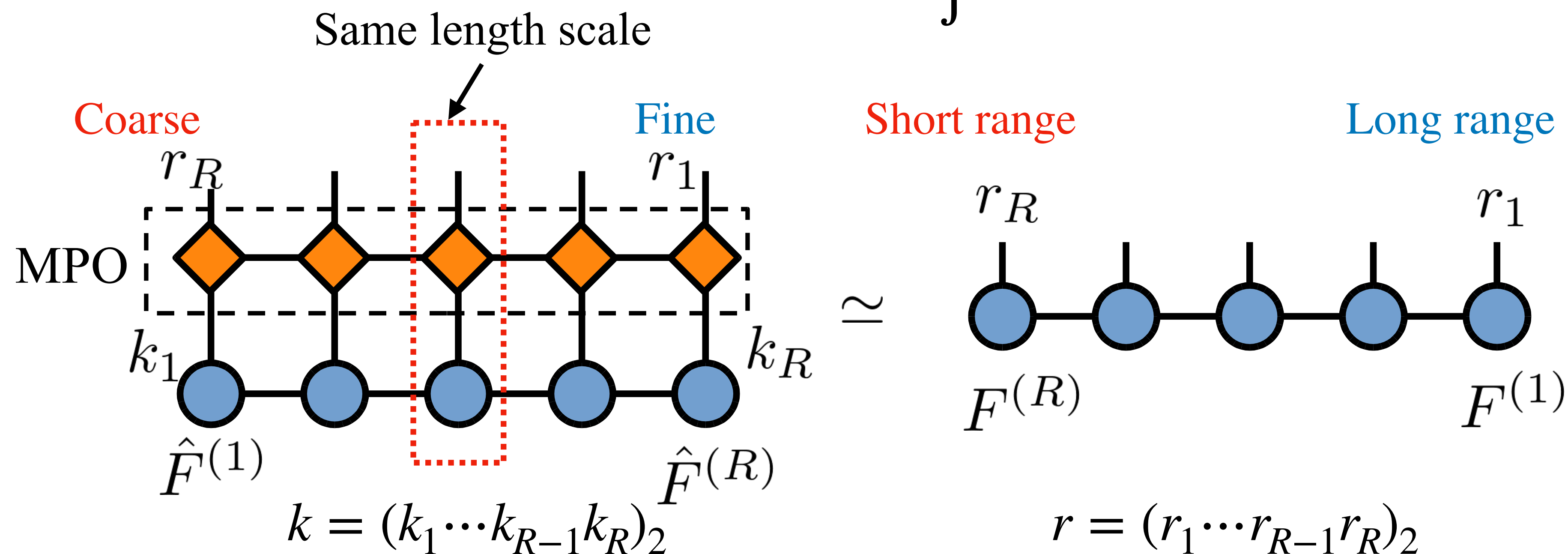
Particle-hole & density channel, compressing **three-frequency** object
 Compression rate $\sim 10^3$
 $|\delta\Gamma|/\max|\Gamma| \sim 10^{-4}$



Computation

Quantum Fourier Transform

$$F(r) = \int dk \hat{F}(k) e^{ikr}$$



Matrix product operator (MPO) can be constructed using quantum Fourier transform or discrete Fourier transform.

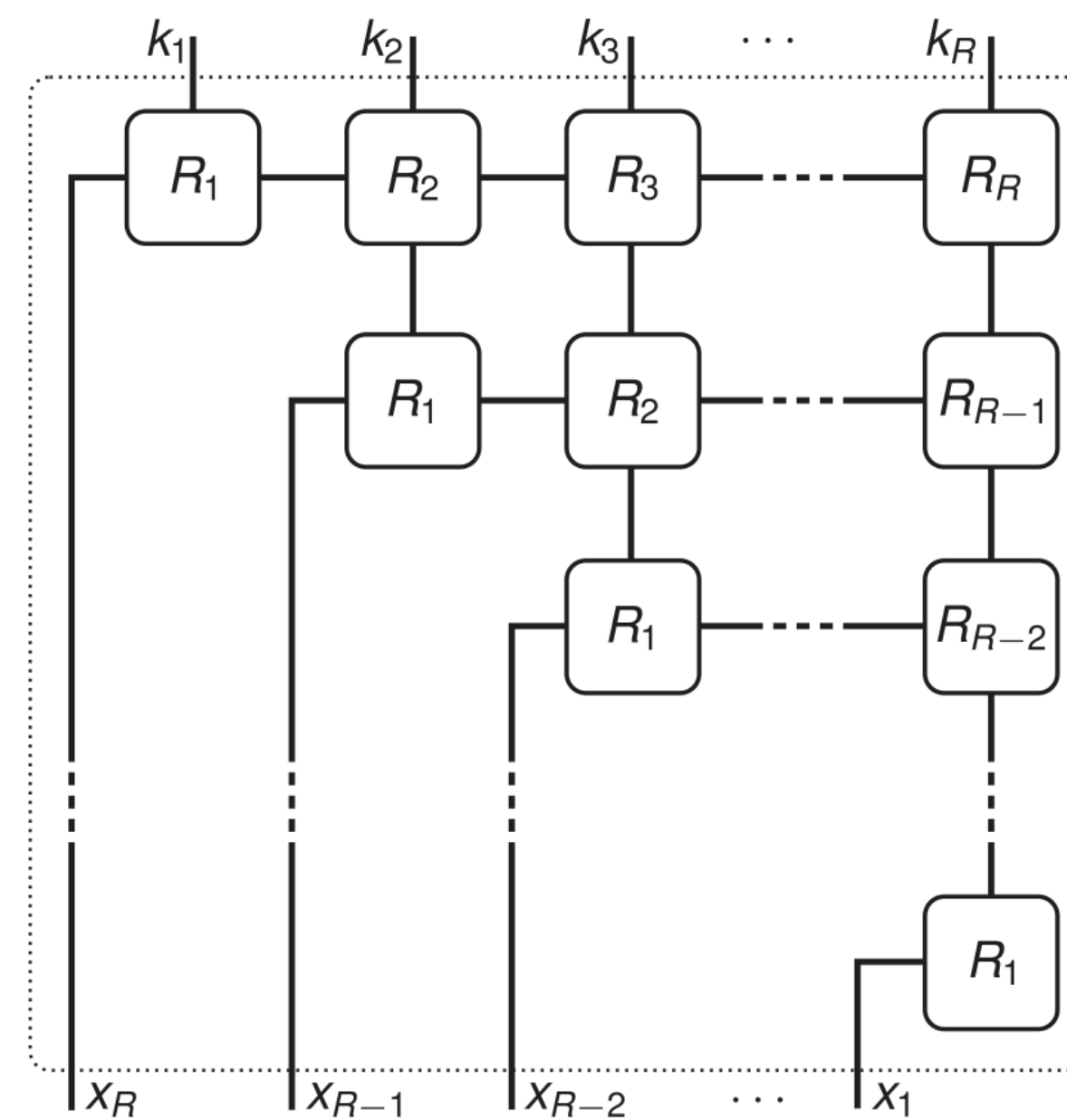
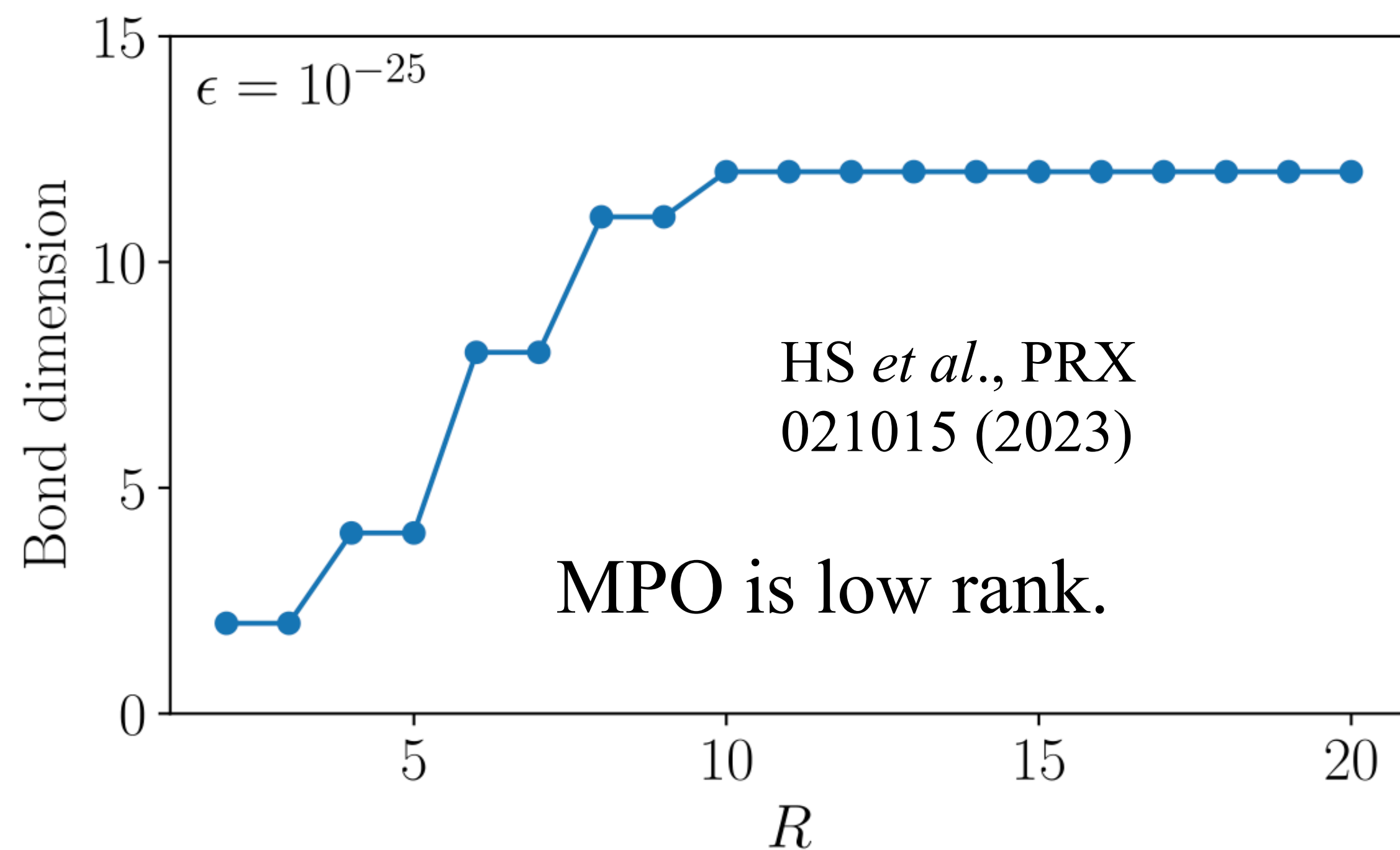
K. J. Woolfe *et al.*, Quantum Inf. Comput. **17**, 1 (2017)

HS *et al.*, PRX 021015 (2023)

J. Chen, E.M. Stoudenmire, S. R. White, arXiv:2210.08468v

Quantum Fourier Transform

$$F(r) = \int dk \hat{F}(k) e^{ikr}$$



Quantum Fourier transform can be simulated on a classical computer efficiently!

∴ Fourier transform is almost independent on different length scales.

K. J. Woolfe *et al.*, Quantum Inf. Comput. **17**, 1 (2017) [only numerical]

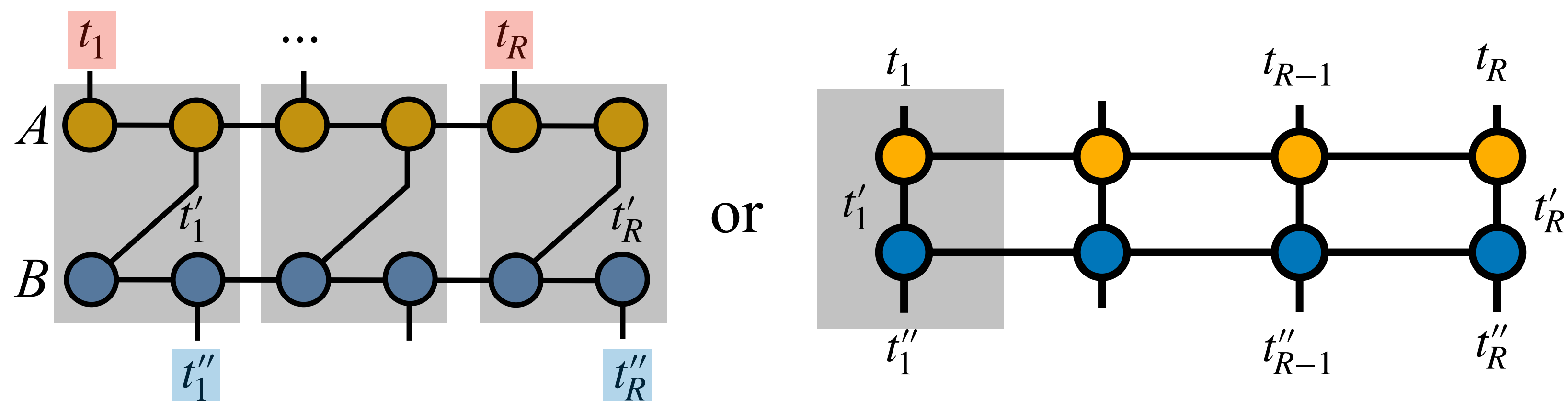
HS *et al.*, PRX 021015 (2023) [only numerical]

J. Chen, E.M. Stoudenmire, S. R. White, arXiv:2210.08468v1 [math proof]

Matrix multiplication/convolution

$$C(t_1, t_1'', \dots, t_R, t_R'') = \sum_{t'_1, \dots, t'_R} A(t_1, t'_1, \dots, t_R, t'_R) B(t'_1, t_1'', \dots, t'_R, t_R'')$$

Tensor network contractions

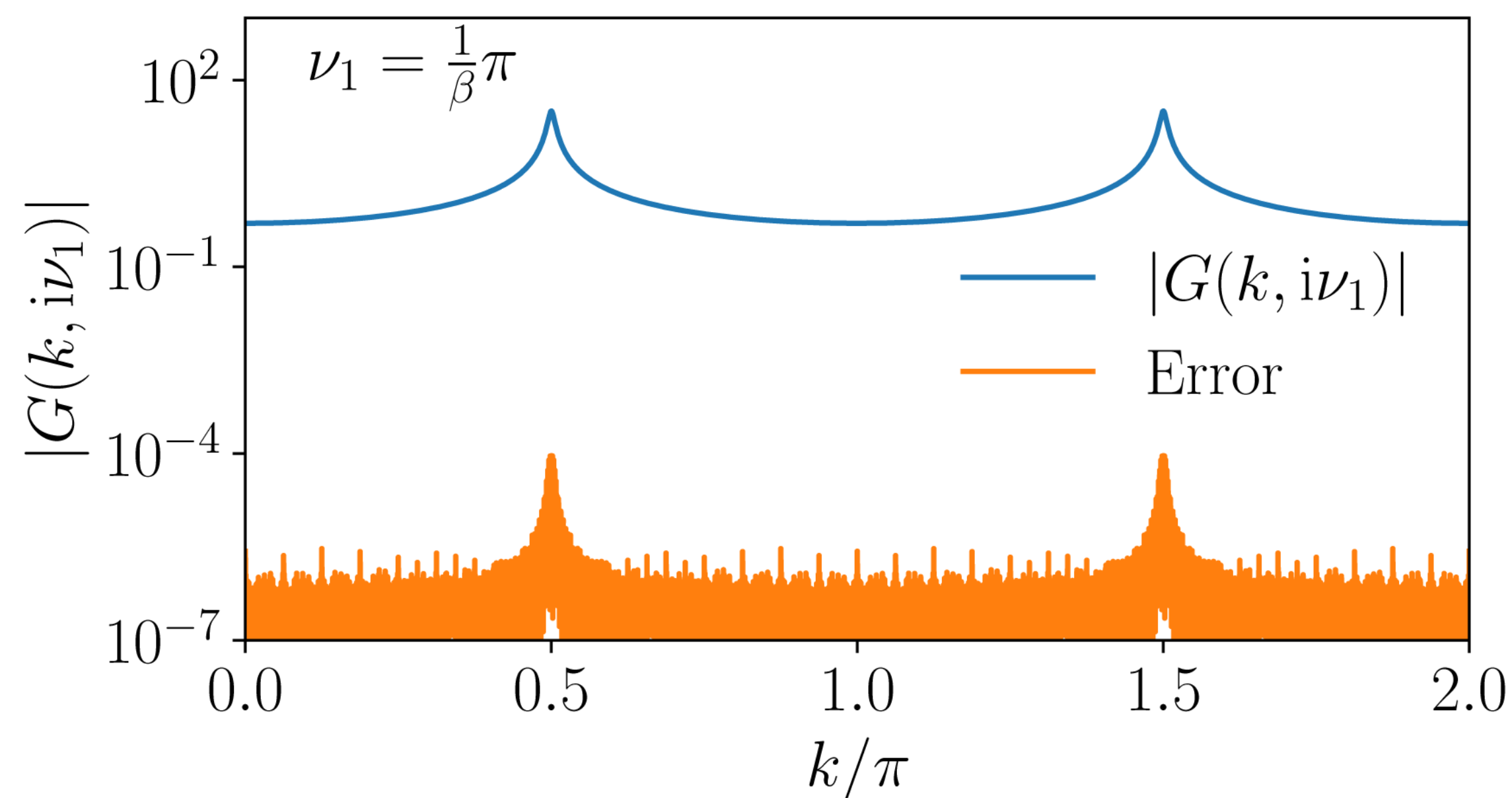
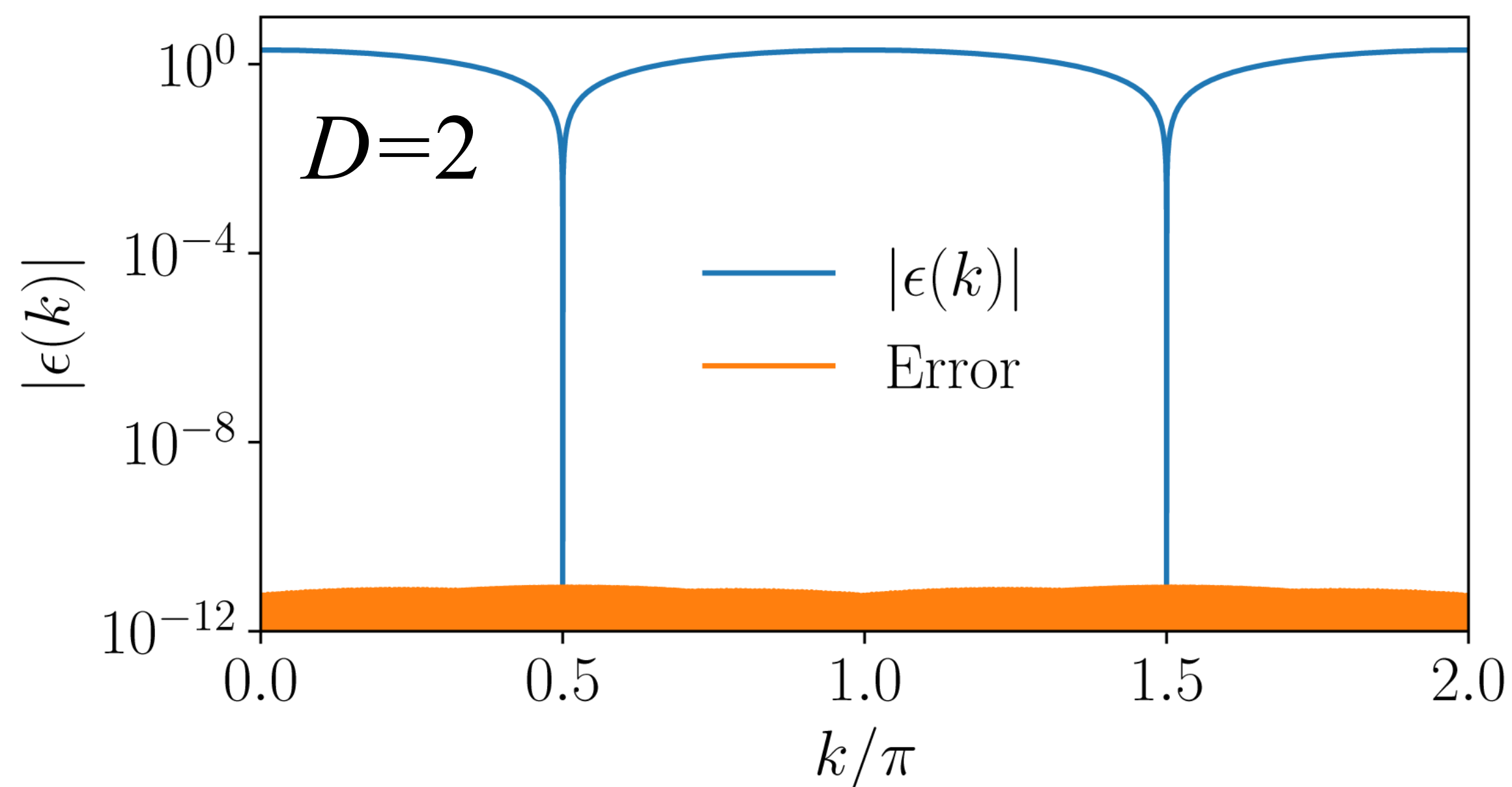
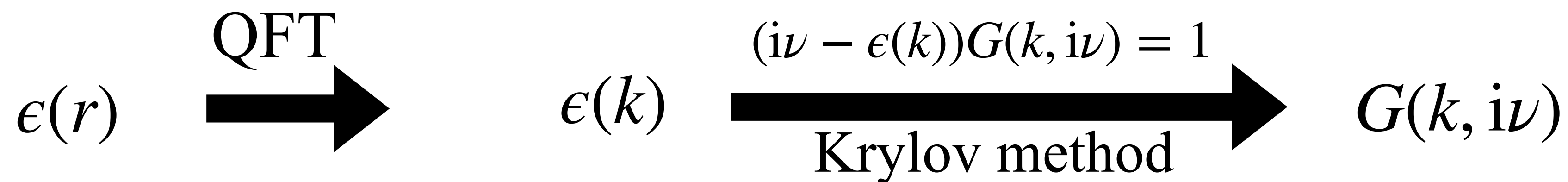


If A , B , and C have a bond dimension of D , the computation time scales as $O(D^4)$.

Dyson equation

1D tight-binding model

$R = 20$ ($2^R \approx 10^6$), $\beta = 100$

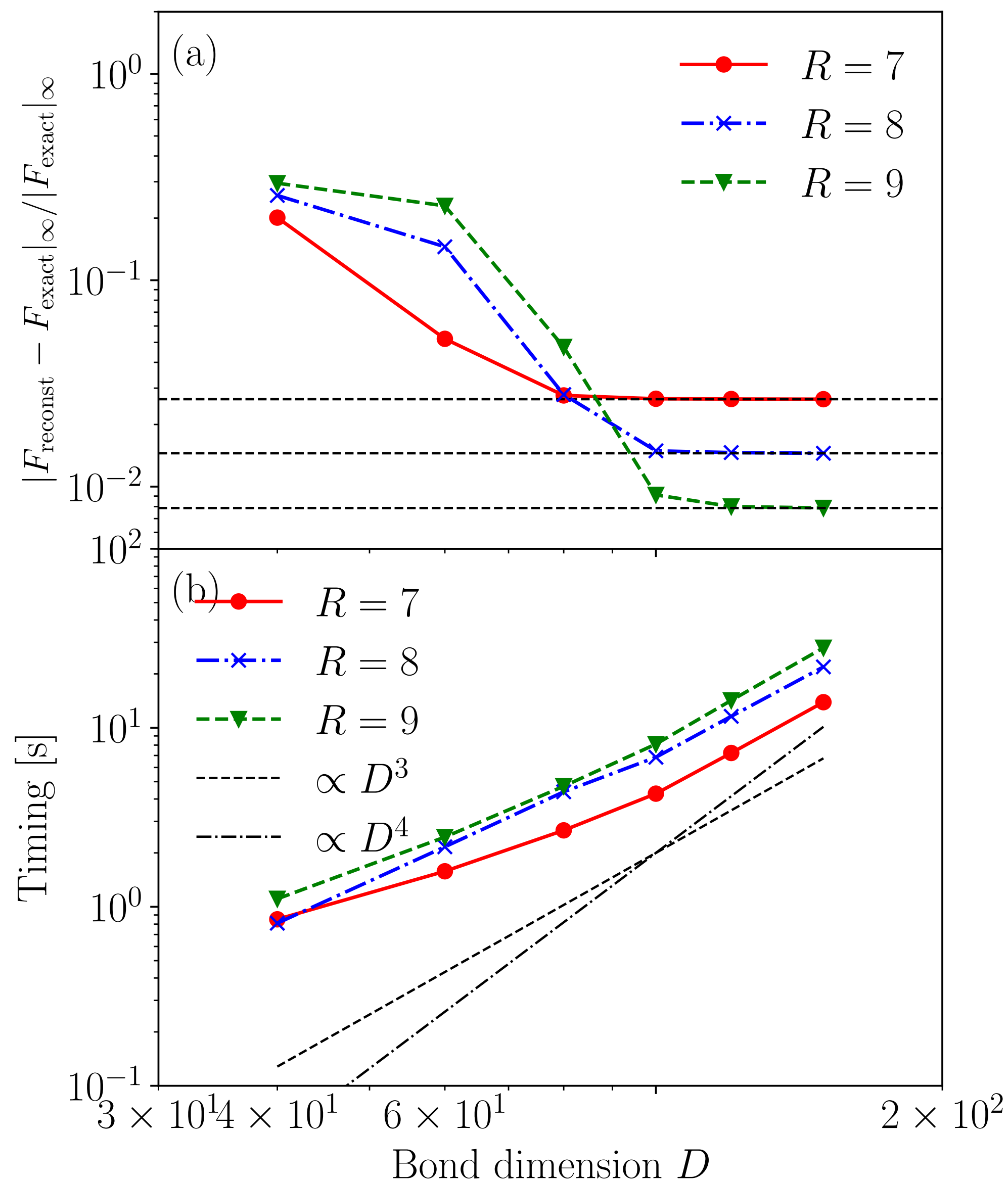


Exponential speed up

In practice, quantics TCI is more efficient (Part II).

Bethe-Salpeter equation

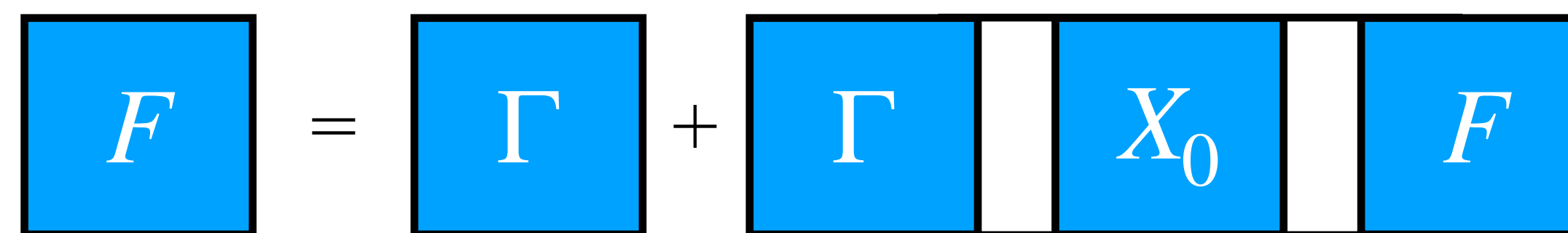
Particle-hole & density channel



Hubbard atom, $U=3$, $\beta=1$

One-shot evaluation of BSE in three-frequency space

$$F_{\text{reconst}} = \Gamma + (\beta^{-2}(\Gamma(X^0 F)))$$



$$F_{d/m}(i\nu, i\nu'; i\omega) = \Gamma_{d/m}(i\nu, i\nu'; i\omega) + \frac{1}{\beta^2} \sum_{\nu'', \nu'''} \Gamma_{d/m}(i\nu, i\nu''; i\omega) X^0(i\nu'', i\nu'''; i\omega) \times F_{d/m}(i\nu''', i\nu'; i\omega),$$

Exponential speed up

Outline

- Introduction
- Part I Quantics tensor train
HS et al., PRX **13**, 021015 (2023)
- Part II Quantics tensor cross interpolation
M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, **HS**, X. Waintal, arXiv:2303.11819
- Outlook & summary

Different routes to quantics

1. Analytic form: exponential
2. SVD of numerical data
3. Computation on the fly: Dyson, Bethe-Salpeter equations.
4. **Quantics Tensor Cross Interpolation**

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, **HS**, X. Waintal, arXiv:2303.11819

Learning a compressed representation from a few function evaluations
Matrix/tensor cross interpolation

Matrix Cross Interpolation (MCI)

Review: N. Kishore Kumar and J. Schneider, Linear Multilinear Alg. **65**, 2212 (2017)

Problem Low-rank approximation of a matrix A by SVD requires knowing **all the matrix elements**....

MCI

$$A \approx CP^{-1}R = \tilde{A}$$

Selected rows

Crosses

Selected columns

$$A = \begin{pmatrix} \bullet & \bullet & \circ & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix} \approx \begin{pmatrix} \circ & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \end{pmatrix} \begin{pmatrix} \circ & \bullet & \bullet \\ \bullet & \circ & \bullet \\ \bullet & \bullet & \circ \end{pmatrix}^{-1} \begin{pmatrix} \bullet & \bullet & \circ & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix},$$

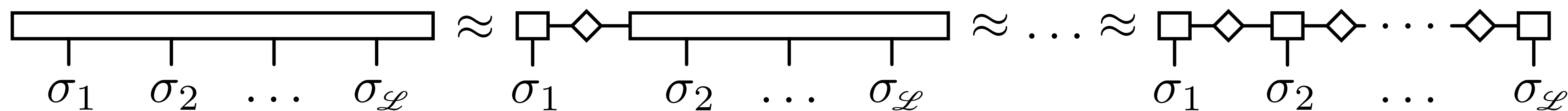
or $\begin{matrix} \square \\ | \\ \square \end{matrix} \approx \begin{matrix} \square & \diamond & \square \\ | & & | \end{matrix} \cdot$

- No need to read all elements of a low-rank matrix!
- Equality is exact on the selected columns and rows.
- Many algorithms to choose good pivots:
e.g., Maxvol (maximize $|P|$)

Tensor cross interpolation (TCI)

I. V. Oseledets, SIAM Journal on Scientific Computing **33**, 2295 (2011)

S. Dolgov and D. Savostyanov, Computer Physics Communications **246**, 106869 (2020)



$$A = \begin{pmatrix} \text{grid of colored dots} \end{pmatrix} \approx \begin{pmatrix} \text{grid of red dots} \end{pmatrix} \begin{pmatrix} \text{grid of purple dots} \end{pmatrix}^{-1} \begin{pmatrix} \text{grid of blue and purple dots} \end{pmatrix},$$

or $\begin{matrix} \square \\ \square \end{matrix} \approx \begin{matrix} \square & \diamond & \square \\ \square & & \square \end{matrix} \cdot$

of function evaluations \propto # of elements in the tensors $\propto O(D^2)$

- Function evaluation on the fly
- No need to store all tensor elements

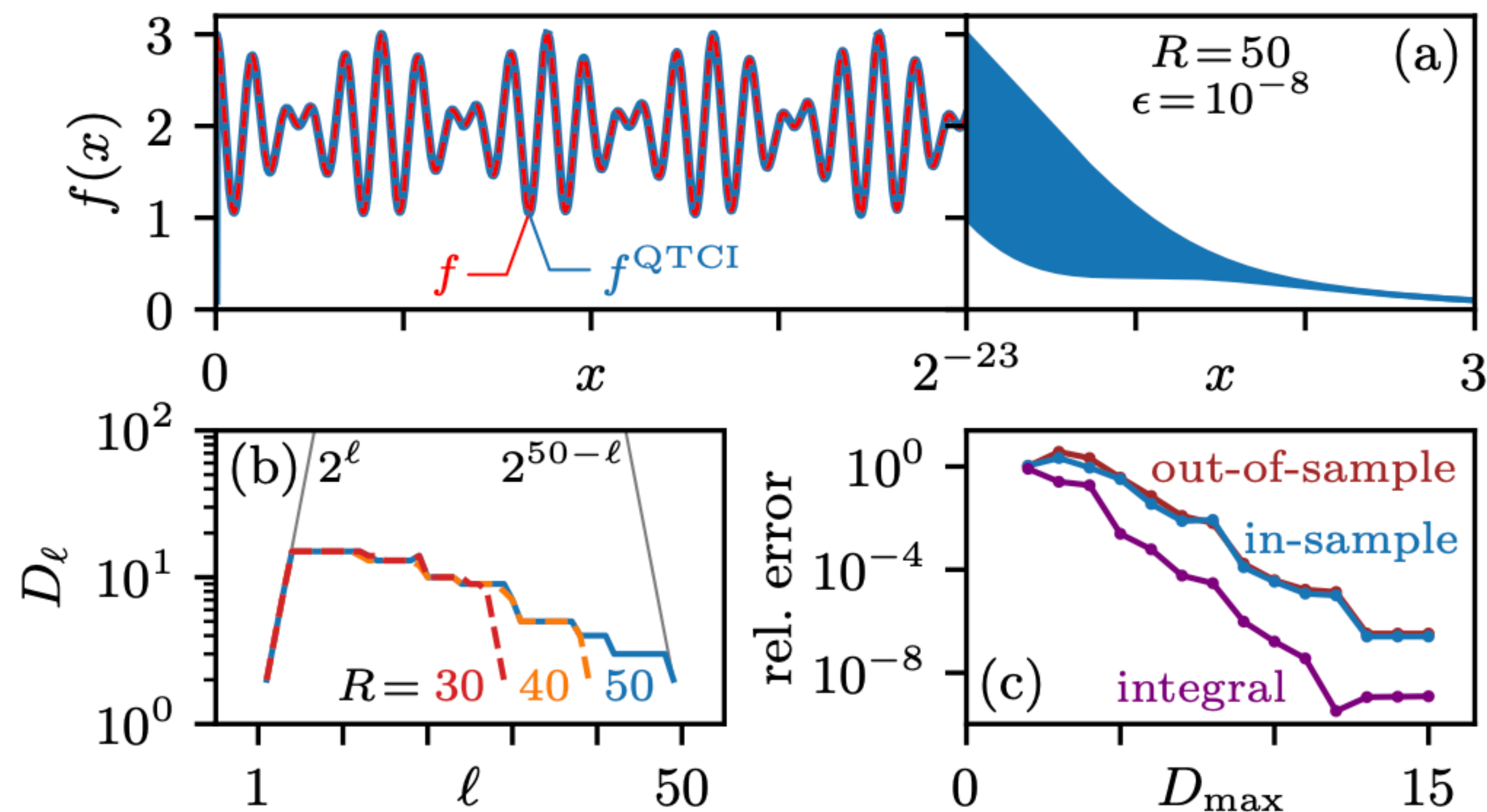
TCI is general and is ready to work with quantics!

1D integration

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, [HS](#), X. Waintal, arXiv:2303.11819

$$f(x) = \cos\left(\frac{x}{B}\right) \underbrace{\cos\left(\frac{x}{4\sqrt{5}B}\right)}_{\text{Fast oscillations}} e^{-x^2} + \underbrace{2e^{-x}}_{\text{Slow decay}}$$

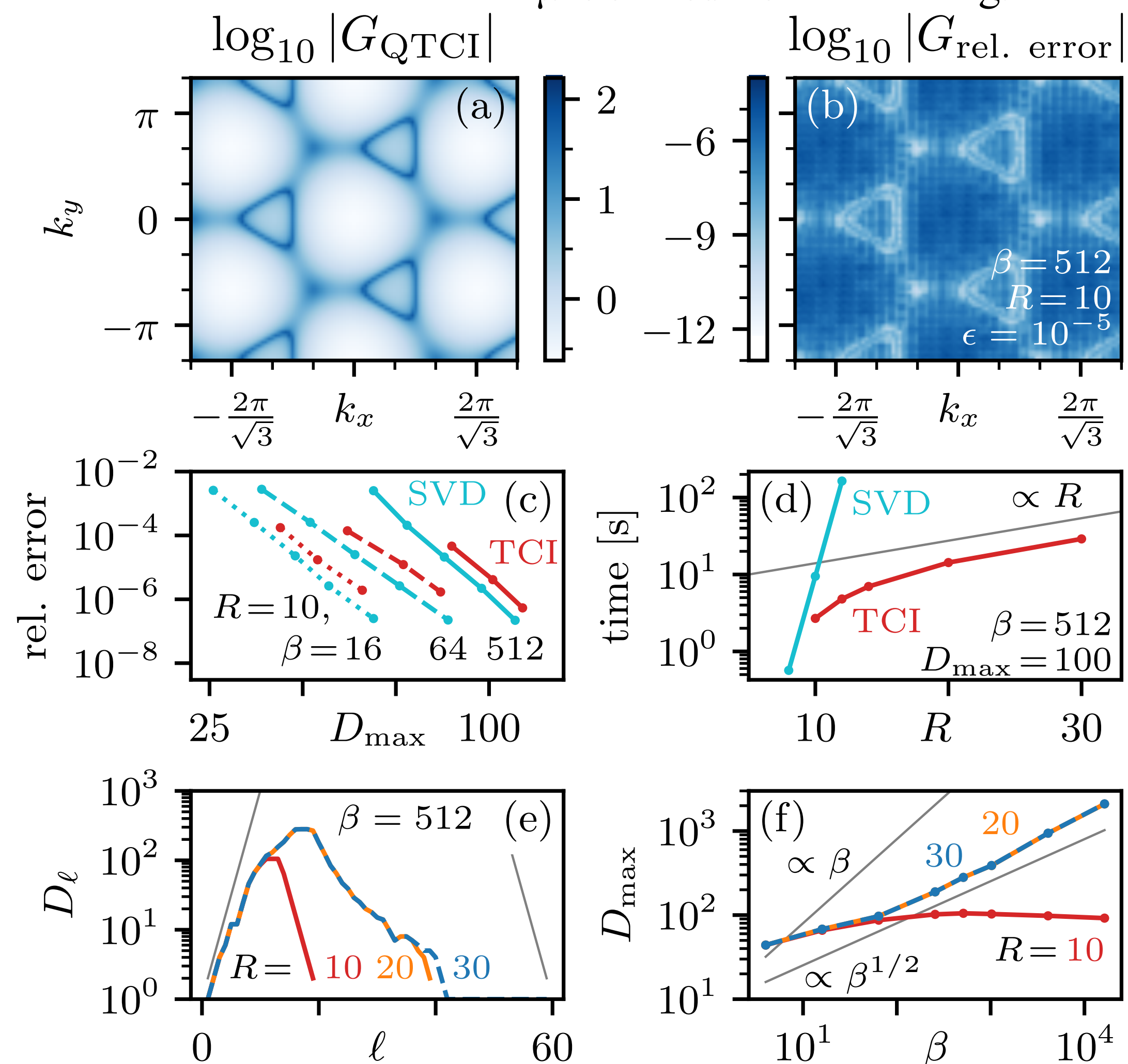
$$\int_0^{\ln(20)} dx f(x) = \frac{19}{10} + O(e^{-1/(4B^2)})$$



Momentum dependence of single-particle Green's function

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, [HS](#), X. Waintal, arXiv:2303.11819

μ is shifted from half filling.



Haldane model

$$H(\vec{k}) = \sum_{i=1}^3 \left[\sigma^1 \cos(\vec{k} \cdot \vec{a}_i) + \sigma^2 \sin(\vec{k} \cdot \vec{a}_i) \right] + \sigma^3 \left[m - 2t_2 \sum_{i=1}^3 \sin(\vec{k} \cdot \vec{b}_i) \right],$$

- Exponential convergence
- QTCI is quasi-optimal (for ranks).
- QTCI is exponentially faster than SVD.
- Bond dimension grows only as $O(\beta^{1/2})$ for 2D.

Chern number

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, [HS](#), X. Waintal, arXiv:2303.11819

Haldane model

$$H(\vec{k}) = \sum_{i=1}^3 \left[\sigma^1 \cos(\vec{k} \cdot \vec{a}_i) + \sigma^2 \sin(\vec{k} \cdot \vec{a}_i) \right] + \sigma^3 \left[m - 2t_2 \sum_{i=1}^3 \sin(\vec{k} \cdot \vec{b}_i) \right],$$

$\delta_m = m - m_c$: deviation from a topological transition

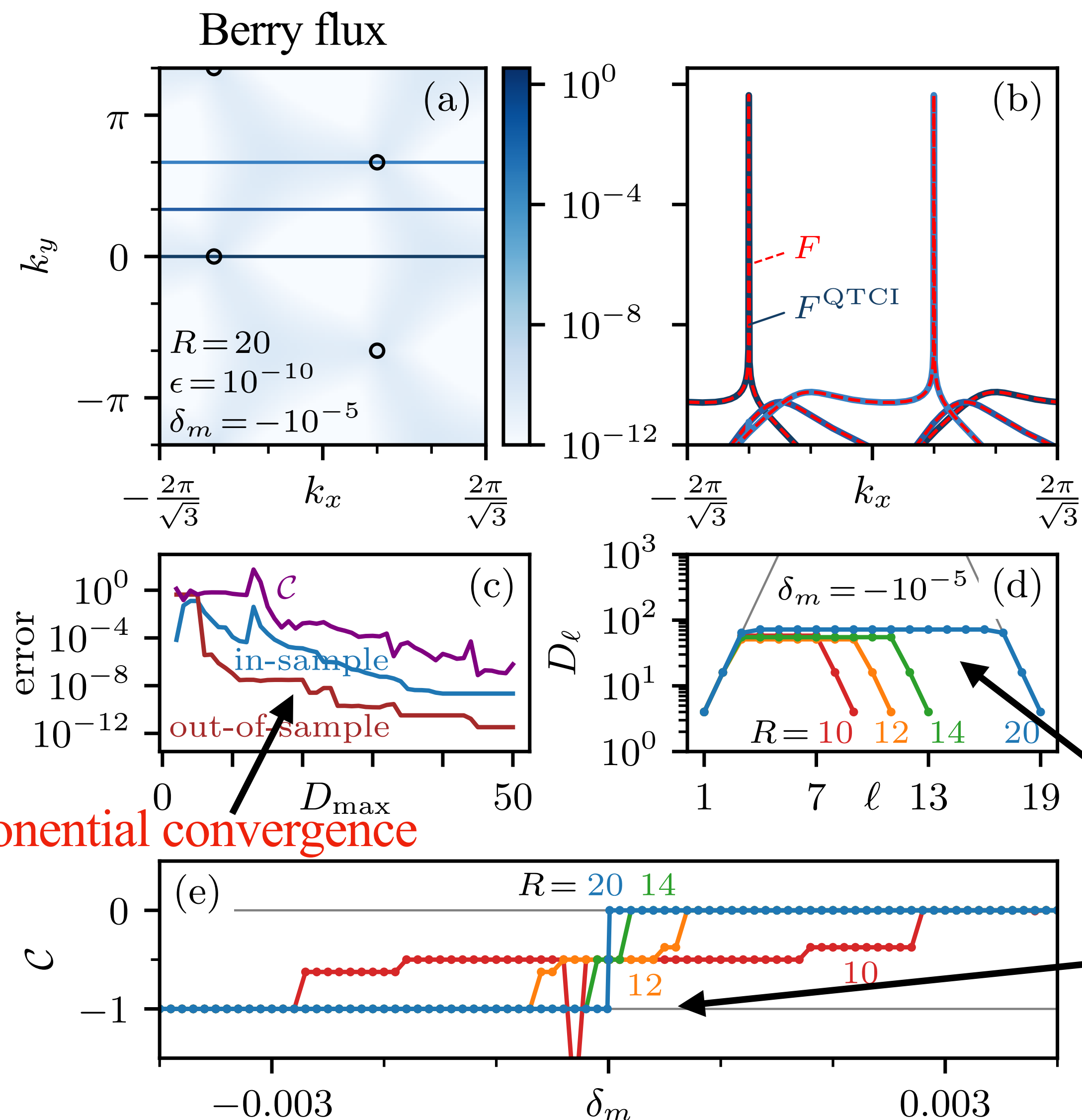
$$C \approx \frac{1}{2\pi i} \sum_{\vec{k} \in \text{BZ}} F(\vec{k})$$

summation of Berry flux over plaquettes

$$F(\vec{k}) \approx -i \arg(\langle \psi_{\vec{k}_1} | \psi_{\vec{k}_2} \rangle \langle \psi_{\vec{k}_2} | \psi_{\vec{k}_3} \rangle \langle \psi_{\vec{k}_3} | \psi_{\vec{k}_4} \rangle \langle \psi_{\vec{k}_4} | \psi_{\vec{k}_1} \rangle)$$

Separation in length scale

Exponentially precise determination
with increasing the number of bits R



Exponential convergence

Outlook

QTT + TCI could be a general framework to attack many problems.

Ab initio calculations

Quantum impurity solvers without sign problem

Diagrammatic calculations at the two-particle level

More general tensor networks

Tree tensor network, PEPS...

Applied math aspects

Comparison with adaptive grids, more robust TCI algorithms

Open-source software



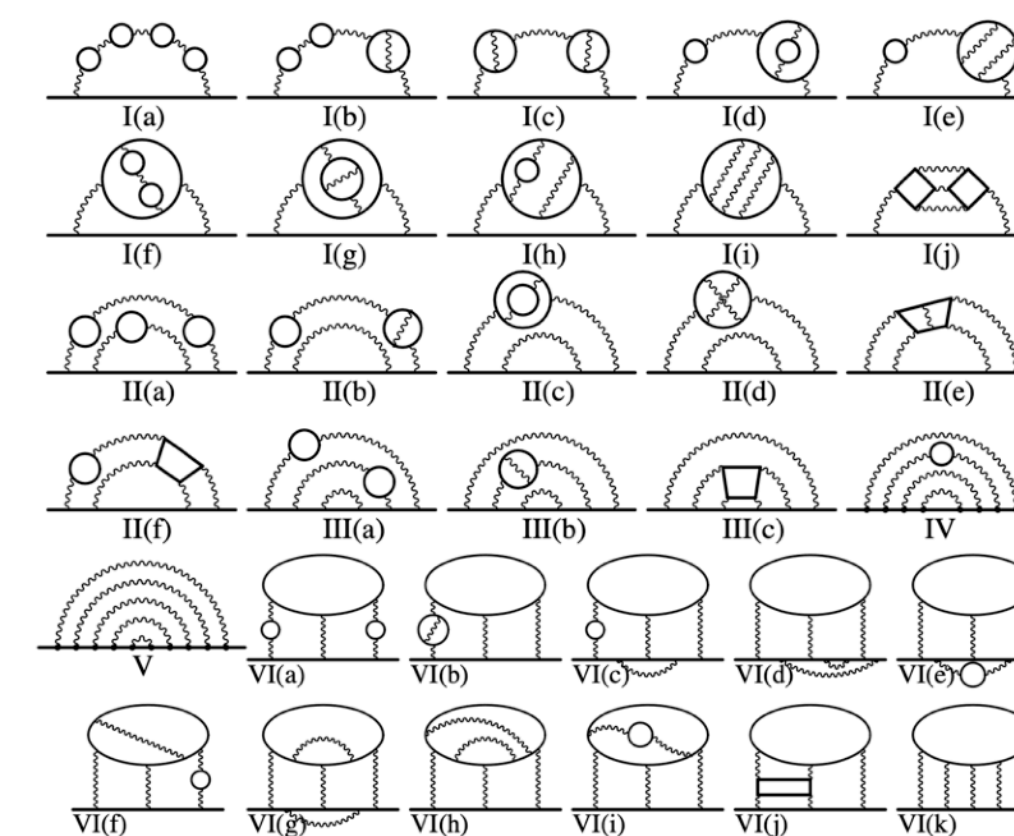
Will be published soon

Beyond condensed matter physics?

Cosmic strings

Planck scale width vs universe-scale length

Perturbation theory in QED

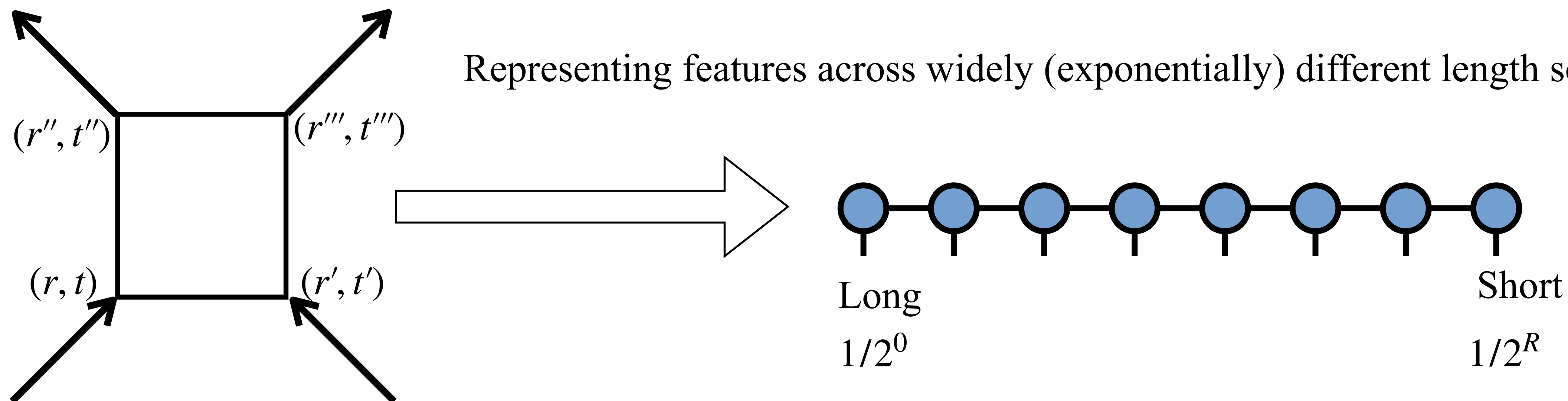


T. Aoyama, T. Kinoshita, M. Nio,
Atoms 7, 28 (2019)

Summary

① Quantics tensor train HS *et al.*, PRX **13**, 021015 (2023)

Representing features across widely (exponentially) different length scales



Computation in compressed form: Fourier transform, convolution...

② Quantics tensor cross interpolation

M. K. Ritter, Y. N. Fernández, M. Wallerberger, J. von Delft, **HS**, X. Waintal,
arXiv:2303.11819

Learning such a representation from a function

Applications: BZ/time integration...