

Progress on generalized approximate contraction

Johnnie Gray
Tensor Subgroup
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High Level Overview:

- Want abstract approach where geometry is just another input:
 - Take **TN geometry**
 - Find **contraction tree**
 - Find **ordering** for tree
 - Choose **gauging** method
 - Choose **compression** method

mix'n'match

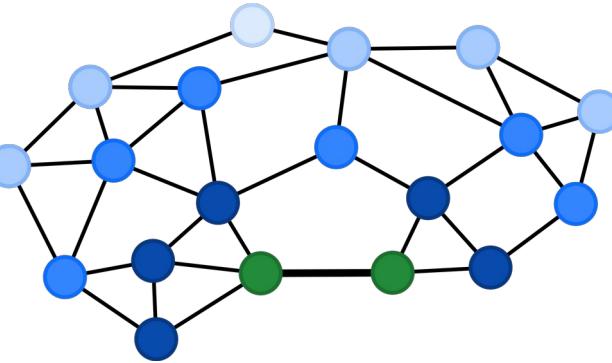
Automatic approximate contraction

Why?

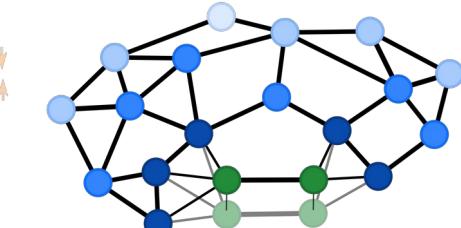
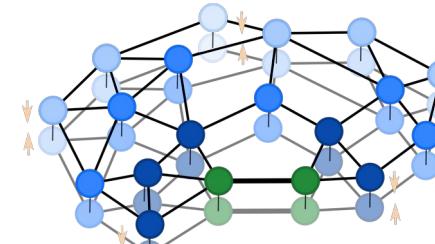
- Unified approach to 2D, 3D, ...
- Contract things with unusual geometries
 - Quantum circuits
 - Weighted model counting instances
 - Weird lattices
 - Partially contracted regular TNs
- Transplant algorithms easily
- Improve existing approaches?

TN Geometry

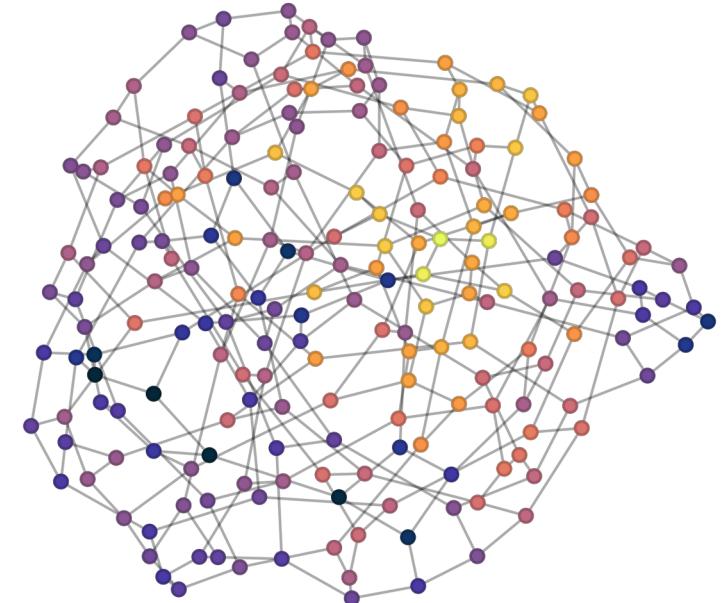
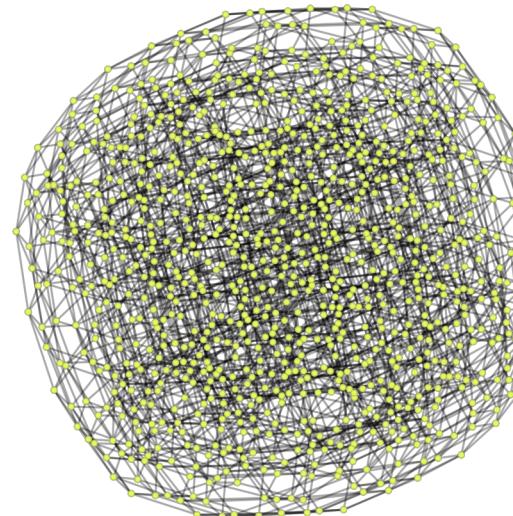
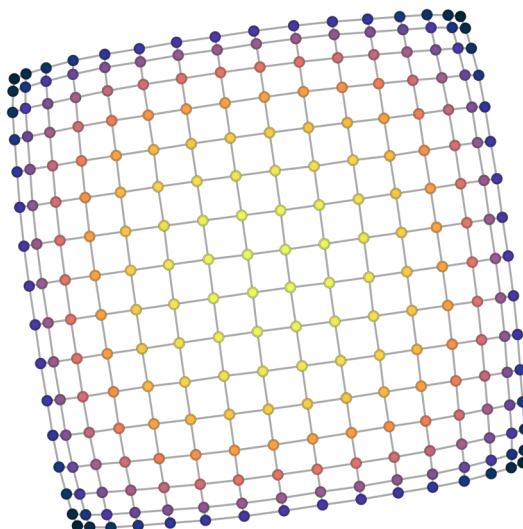
Scalar TN:



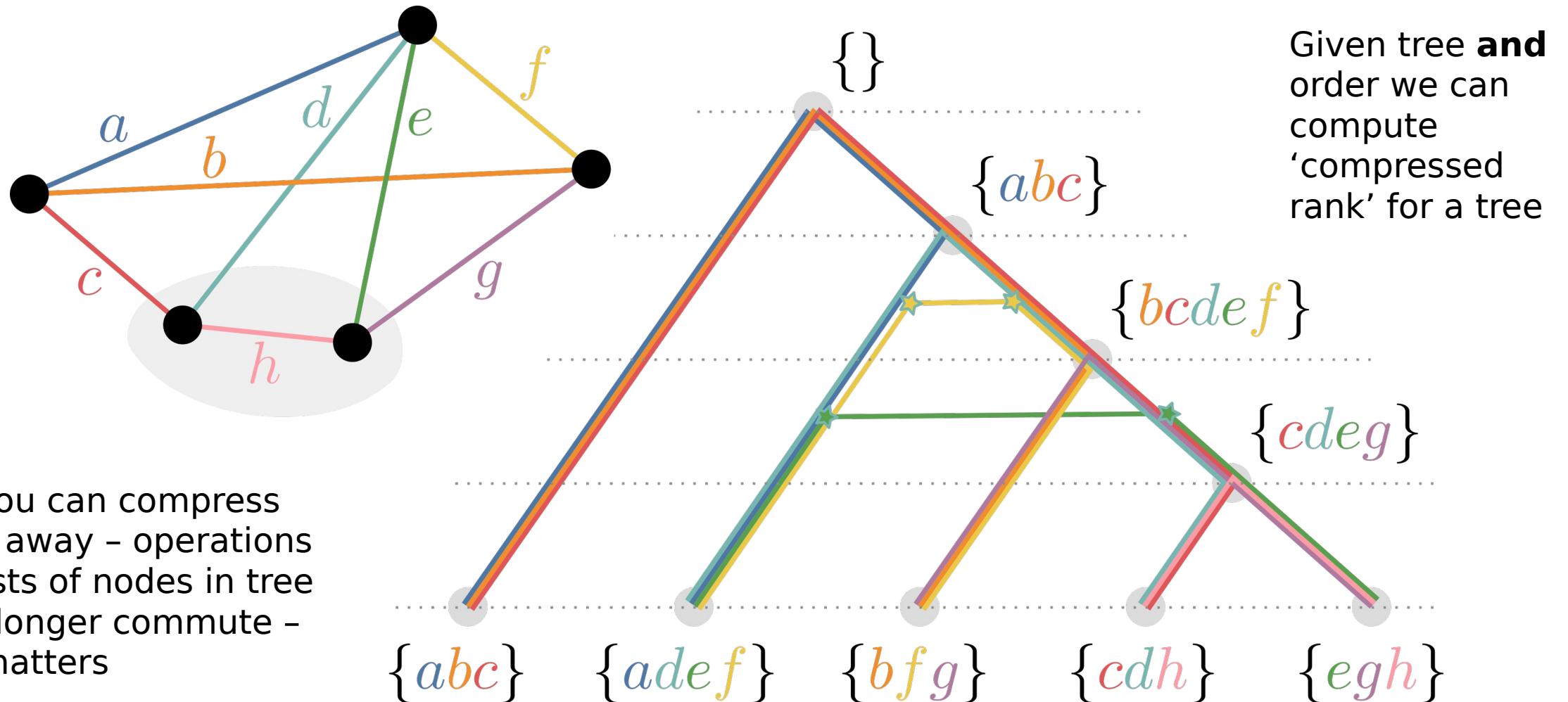
'vector' TN:



Centrality –
useful
heuristic

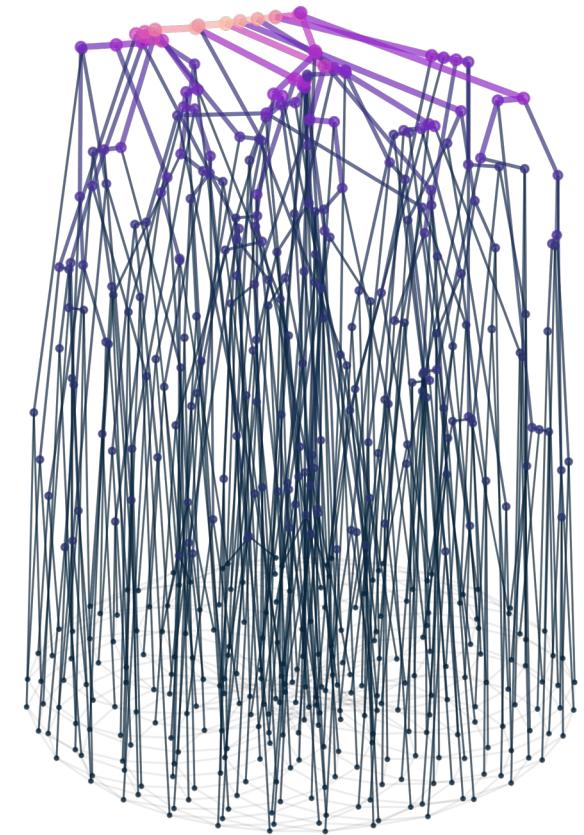


Tree Ordering & ‘Compressed rank’



TN Contraction Tree - Compressed Greedy

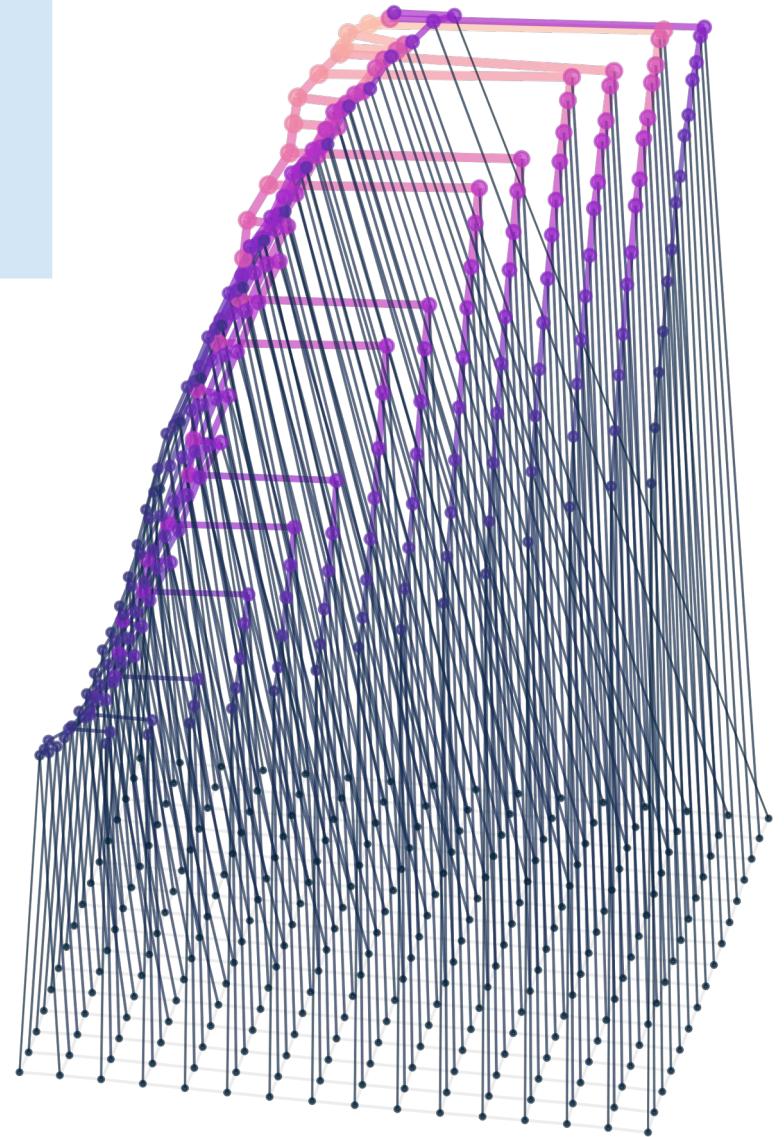
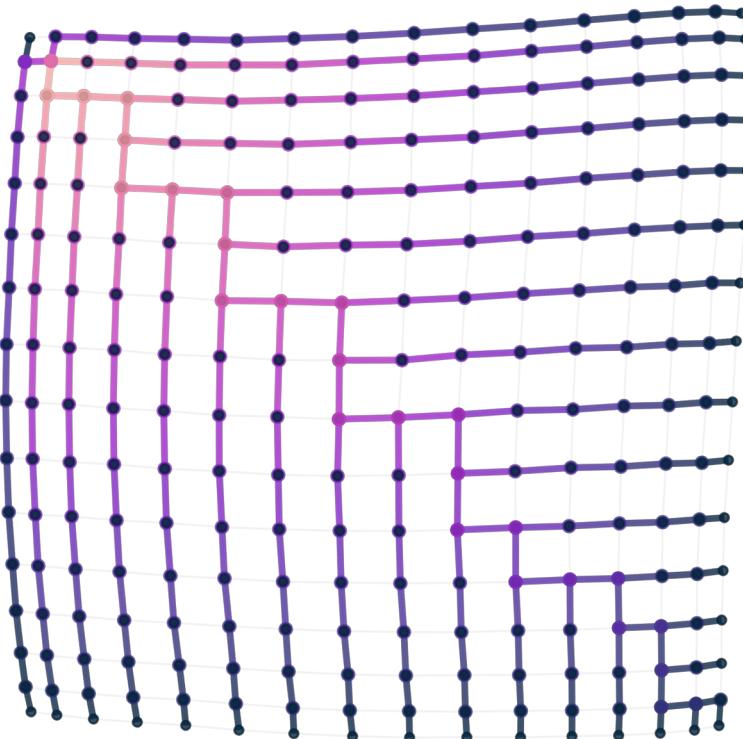
- Generic approach, assess all possible contractions, and greedily select based on
 1. Compressed output size
 2. Centrality
 3. Subgraph size
 4. ...
- Take hyper-optimization approach - tune heuristic function that assesses contraction with respect to total compressed rank



TN Contraction Tree – ‘Boundary Style’

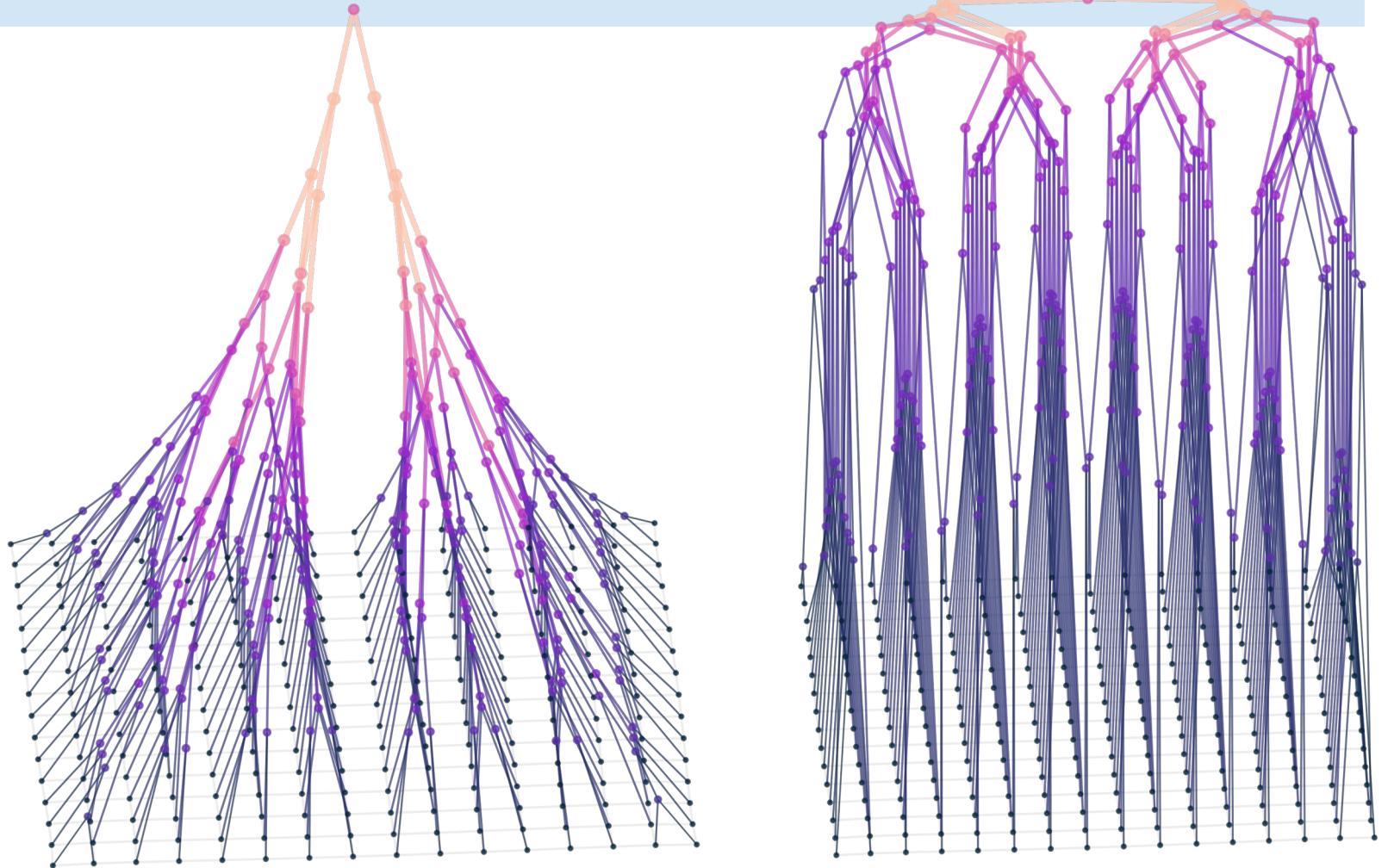
Generate tree from
span of original graph:

- Feed a few heuristics like centrality into span generator, then check quality and repeat
- Restricted set of contraction trees, works well for regular lattices

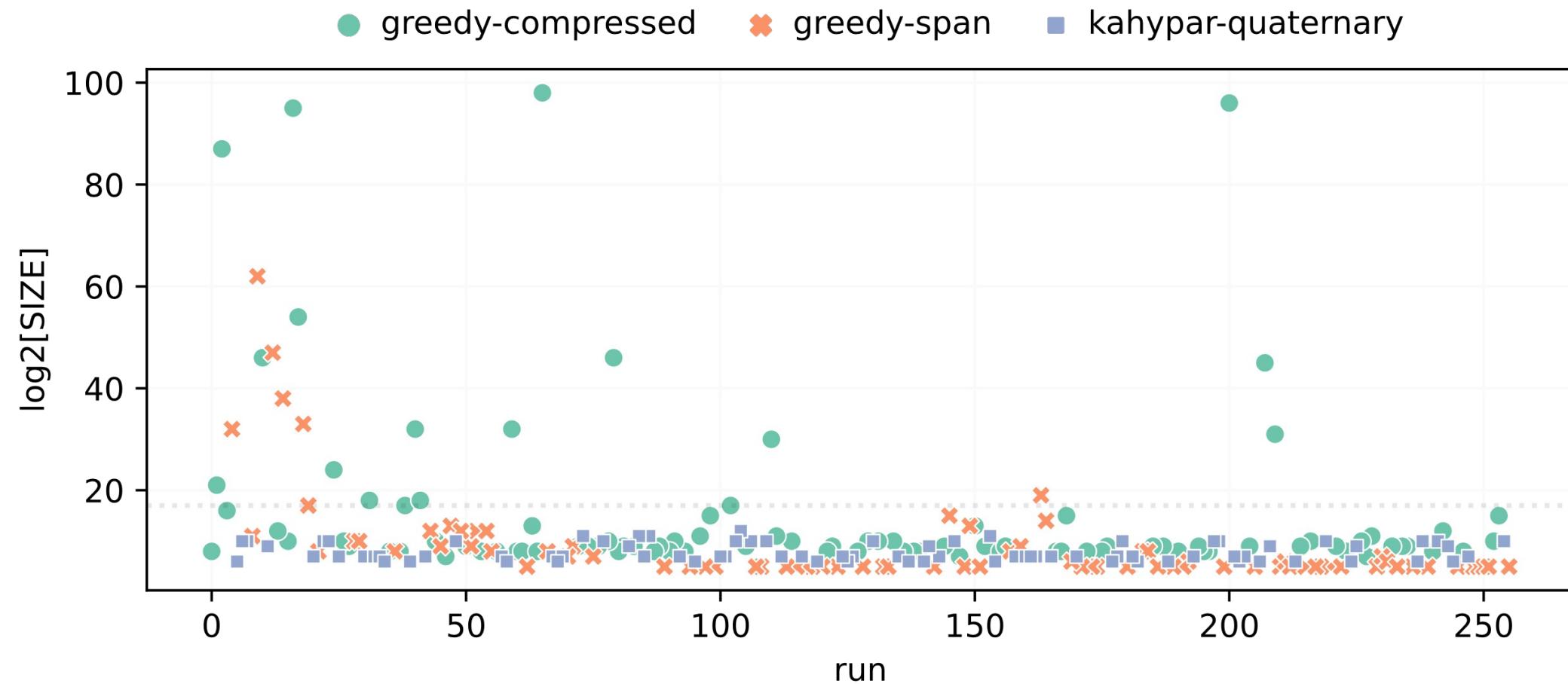


TN Contraction Tree – ‘RG-style’

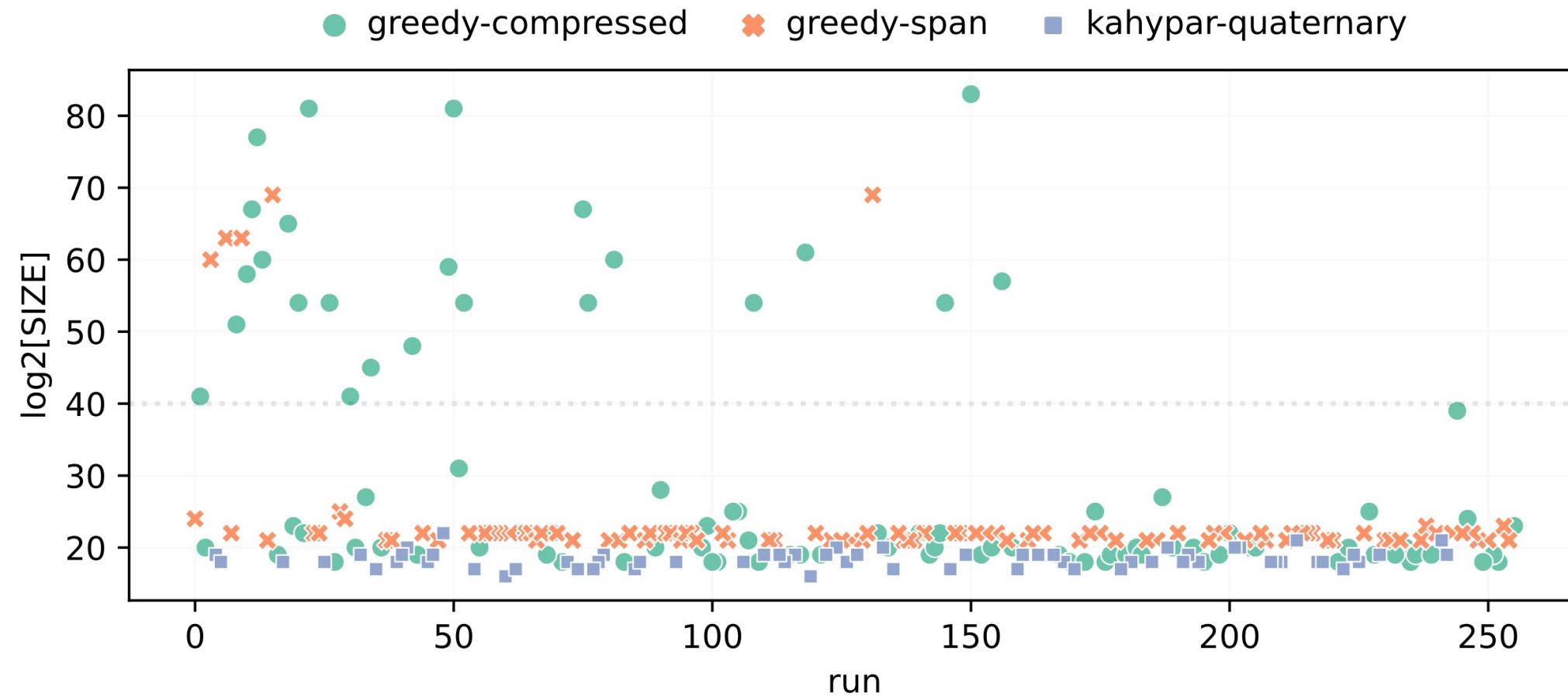
- Generate tree by grouping together small communities of tensors, compressing between these essentially coarse grains the lattice
- Seems good for cases where no obvious boundary or regularity



Compressed rank search results - lattice



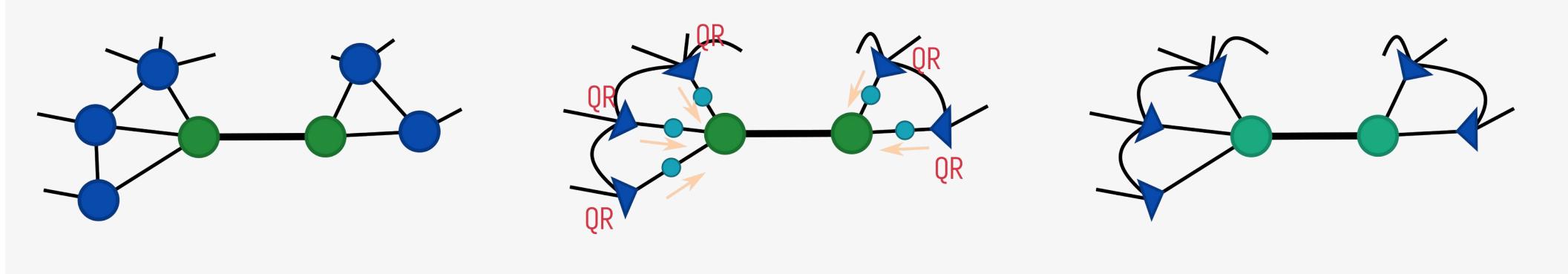
Compressed rank search results - random



Gauging methods – local ‘canonization’

- Basic and cheap strategy that would exactly orthogonalize any tree tensor network is the distance is set to .

local canonization "perfect for trees"

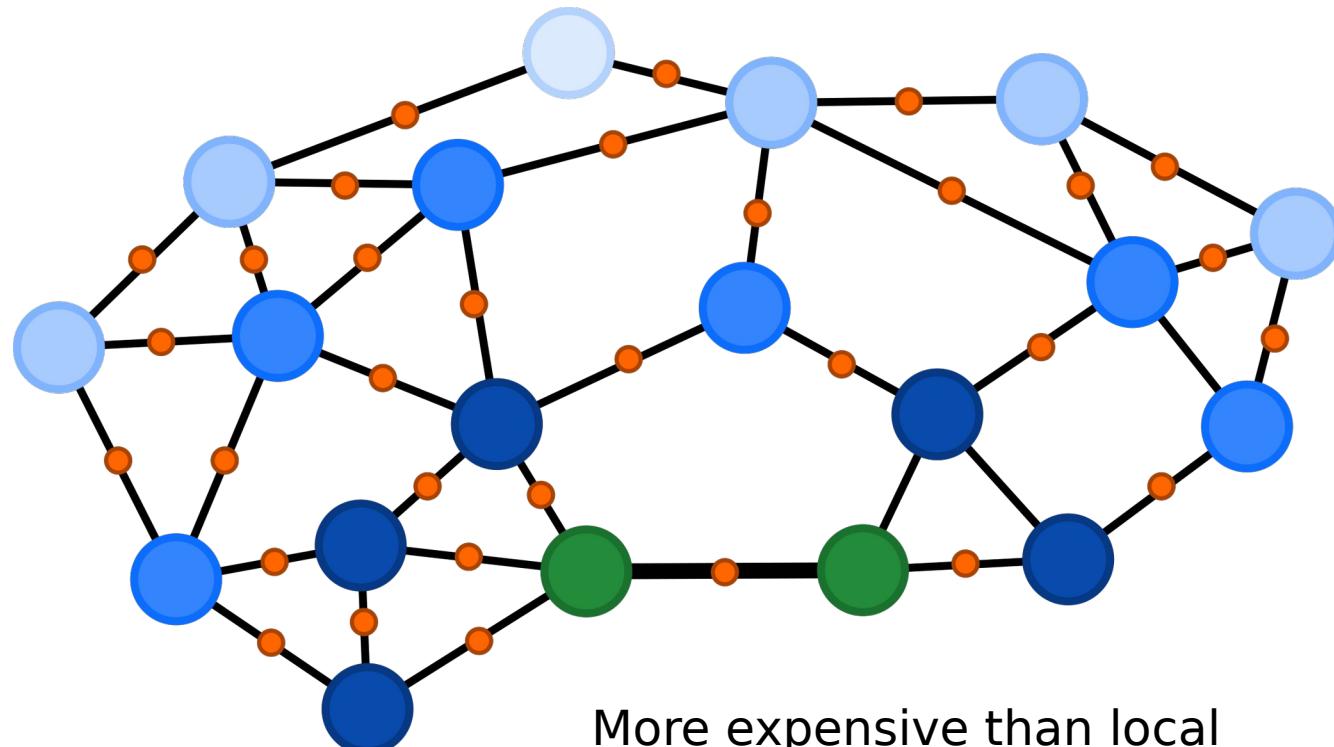


Gauging methods – ‘simple’ gauging

1. Iteratively perform reduced SVD on all bonds and accumulate singular values

2. Absorb (maybe not?)

3. Can re-do for just local portions on TN e.g. after contraction or



More expensive than local canonization but overhead still constant

Gauging methods – ‘full-bond’ gauging

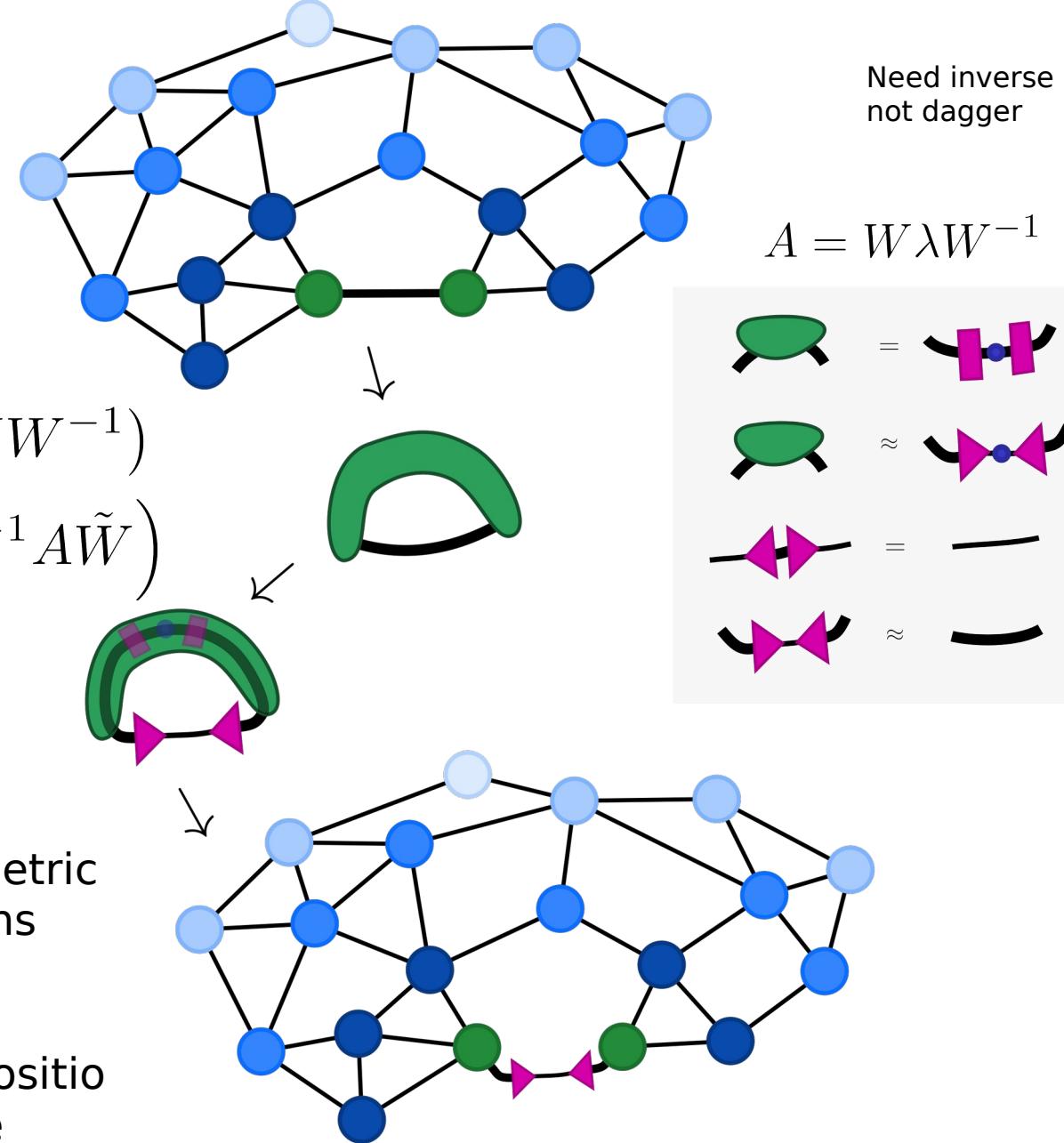
What if we can (approximately) contract the full environment of a bond?

Want to insert a pair of isometries that *would* act as gauges if full rank, that preserve the trace value

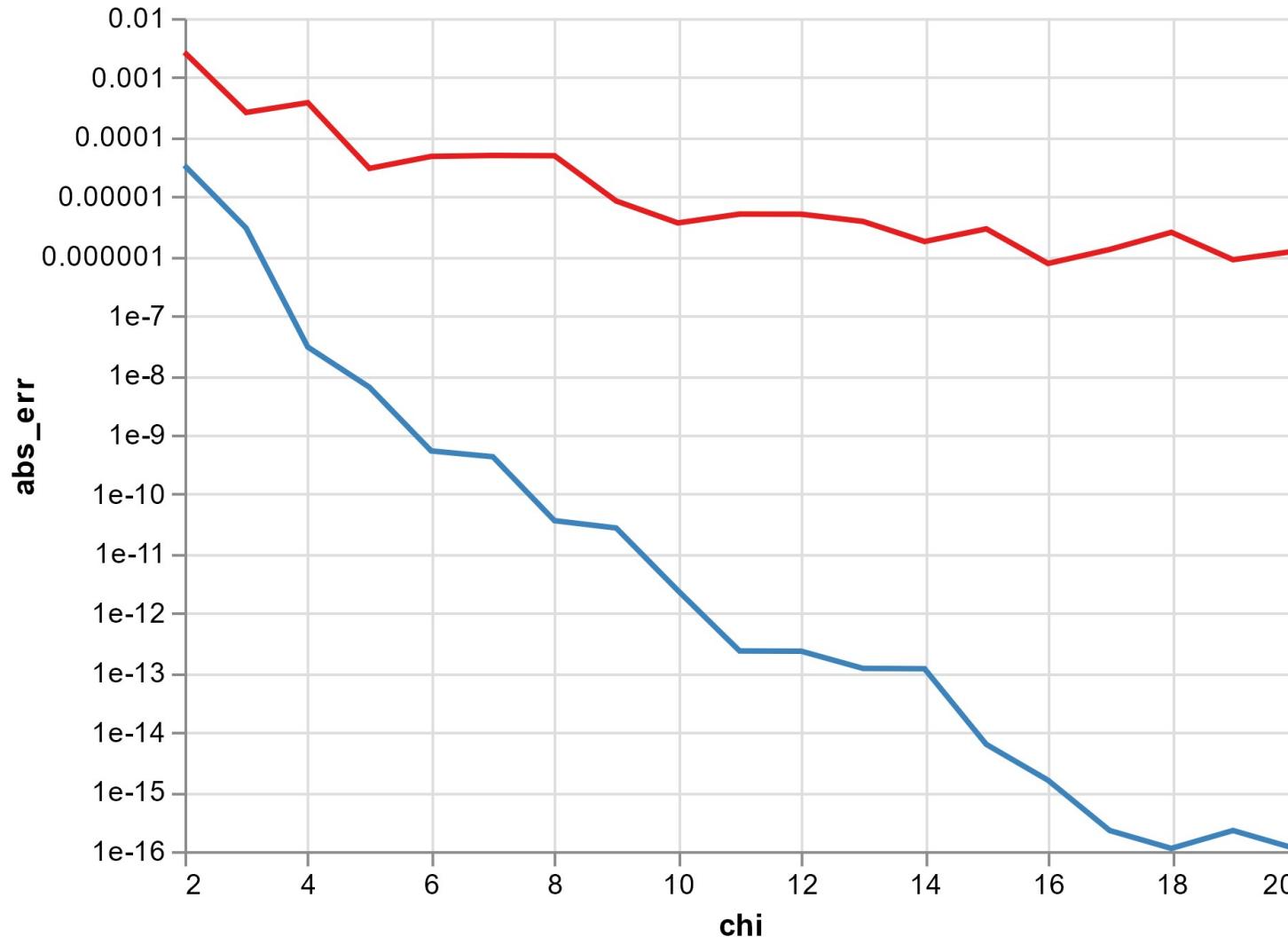
Trace value is not guaranteed to be positive

$$\begin{aligned} z &= \text{Tr}(A) \\ &= \text{Tr}(AWW^{-1}) \\ &\approx \text{Tr}(\tilde{W}^{-1}A\tilde{W}) \end{aligned}$$

Various symmetric decompositions possible, non-Hermitian eigendecomposition basic choice



Results – 16x16 Ising lattice $\beta=0.44$ - [easy]

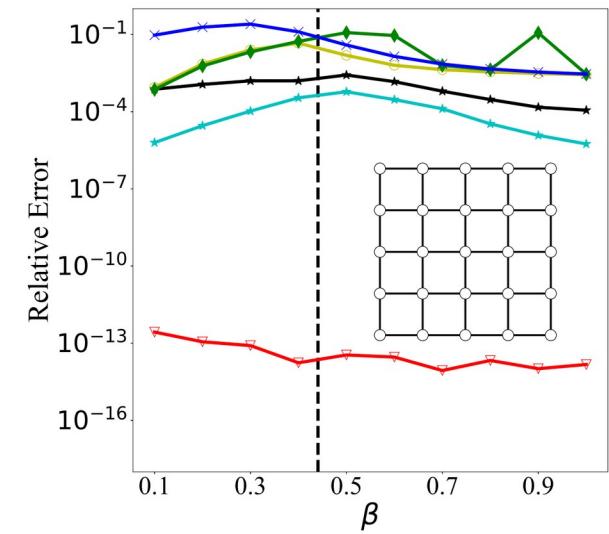


canonize
— 0
— 1

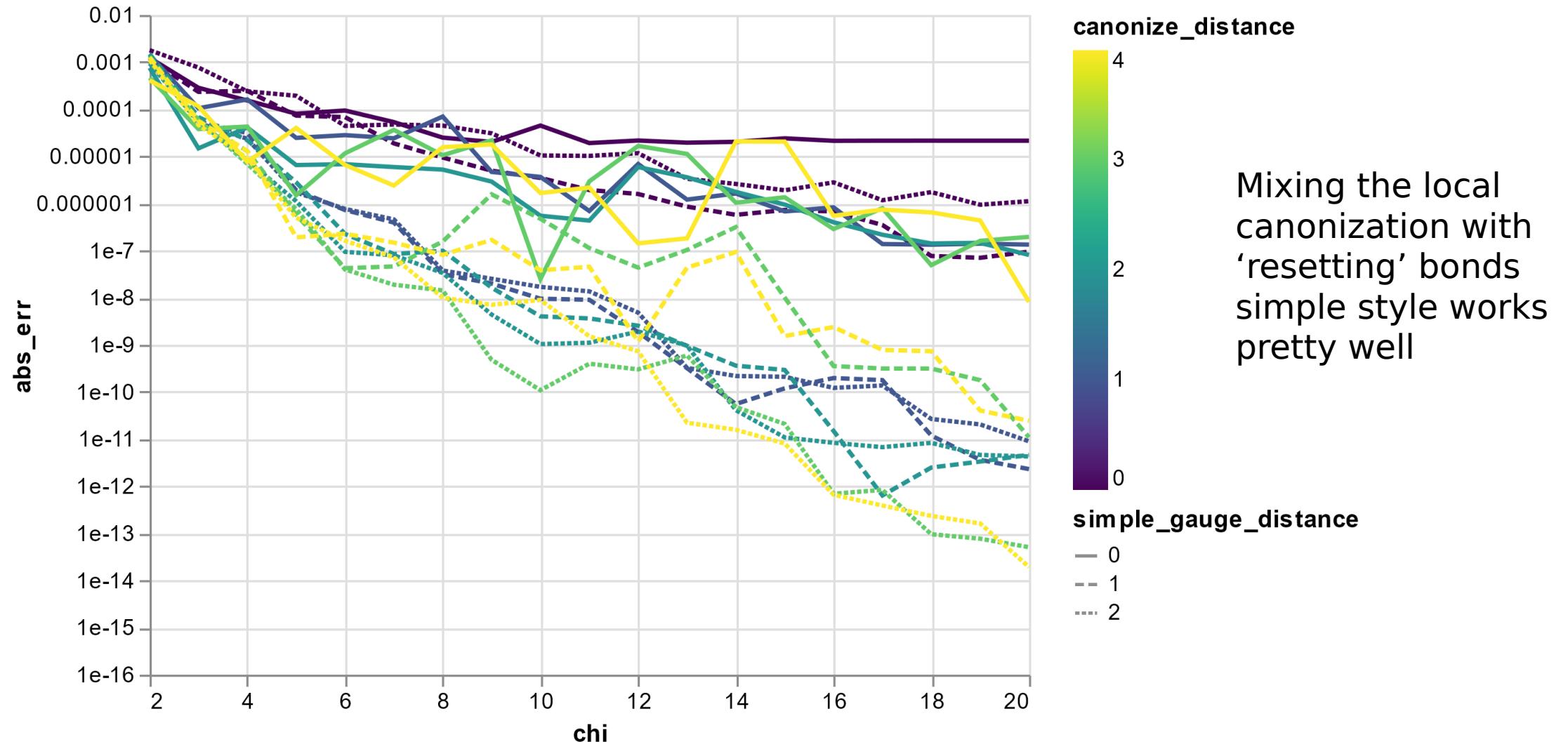
Normal MPS
boundary
contraction already
overpowered

Contracting Arbitrary Tensor Networks: General Approximate Algorithm and Applications in Graphical Models and Quantum Circuit Simulations

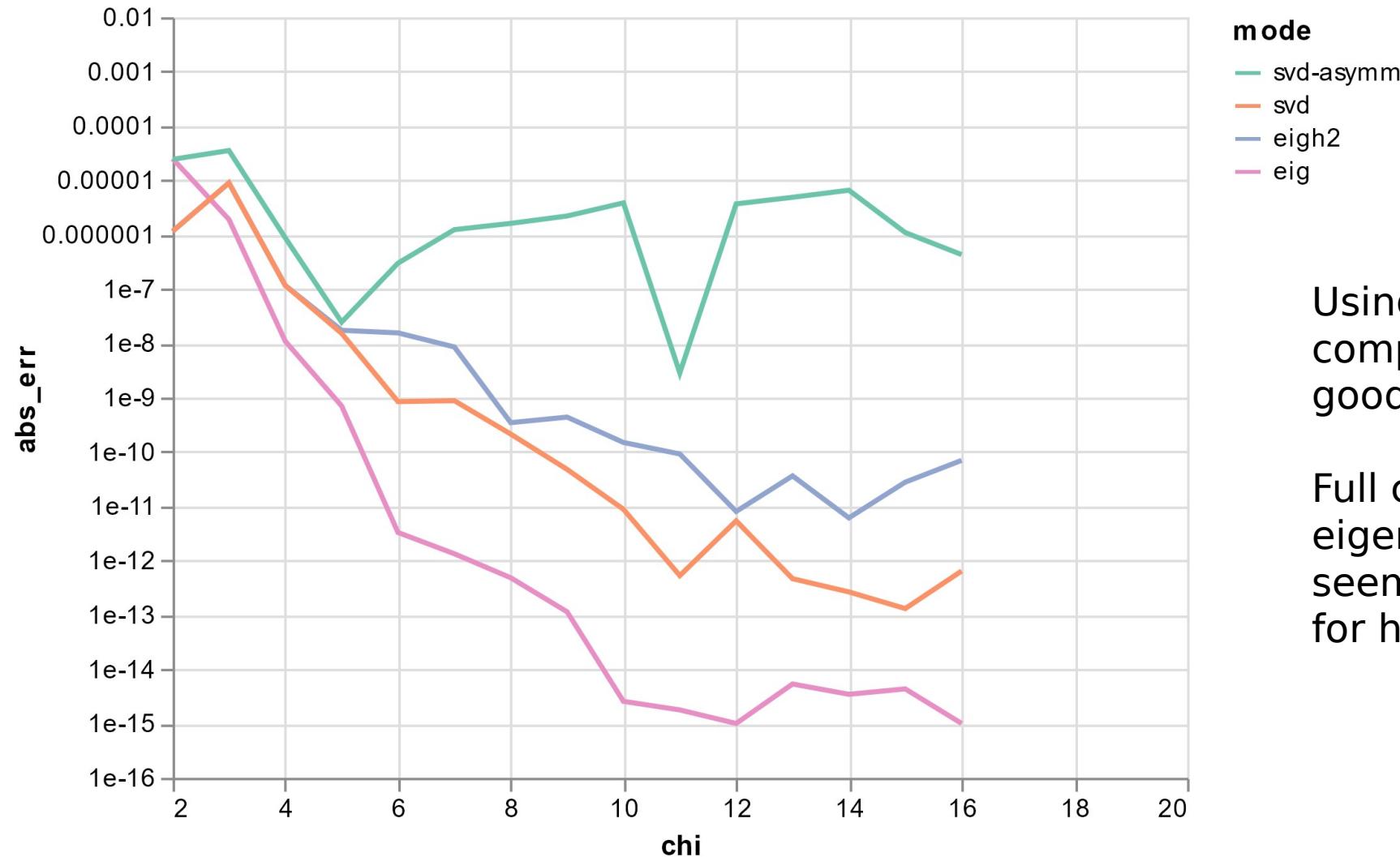
Feng Pan,^{1,2,*} Pengfei Zhou,^{1,2,*} Sujie Li,^{1,2,*} and Pan Zhang^{1,3,4,†}



Results – 16x16 Ising lattice $\beta=0.44$ - [easy]



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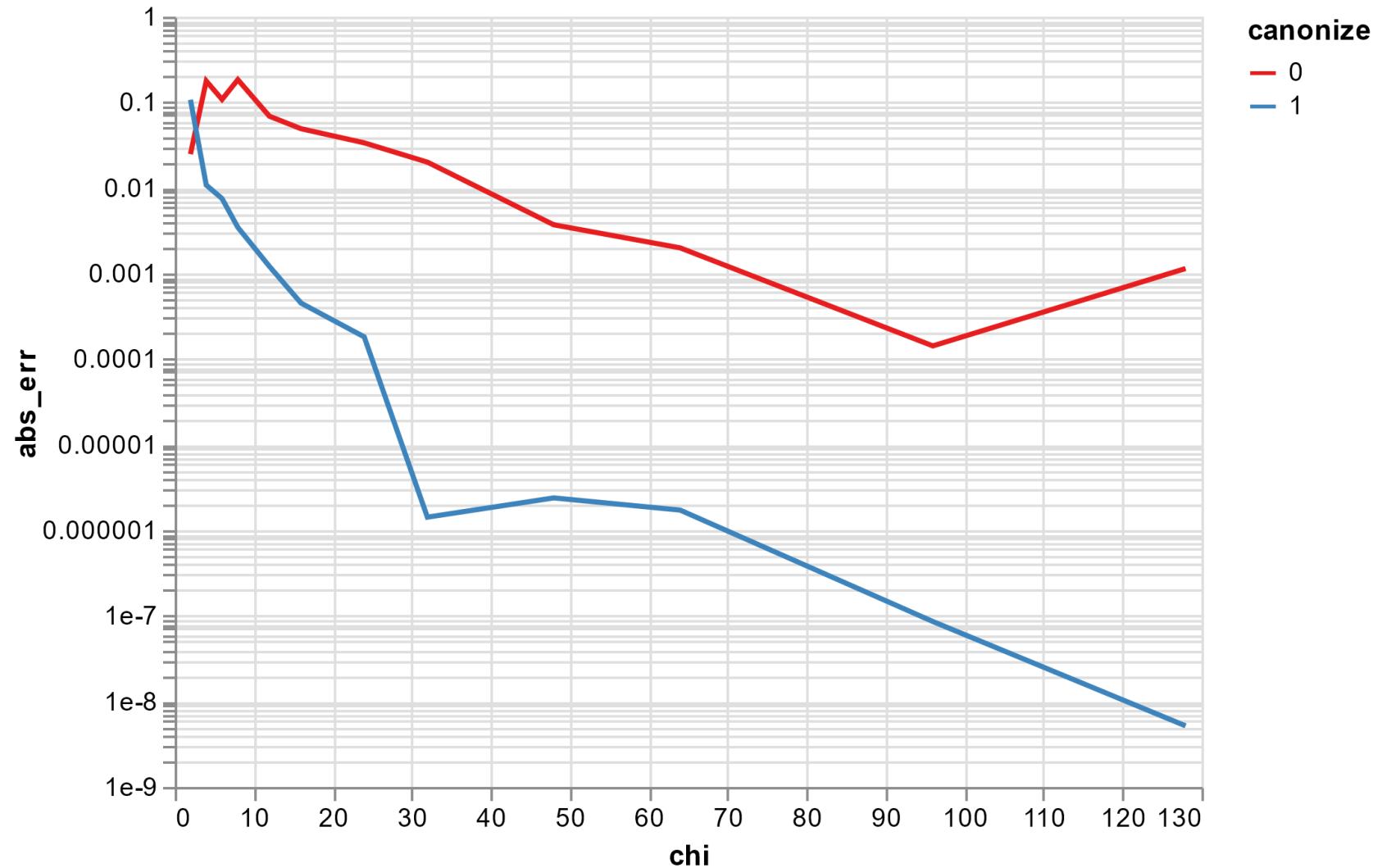
mode

- svd-asymm
- svd
- eigh2
- eig

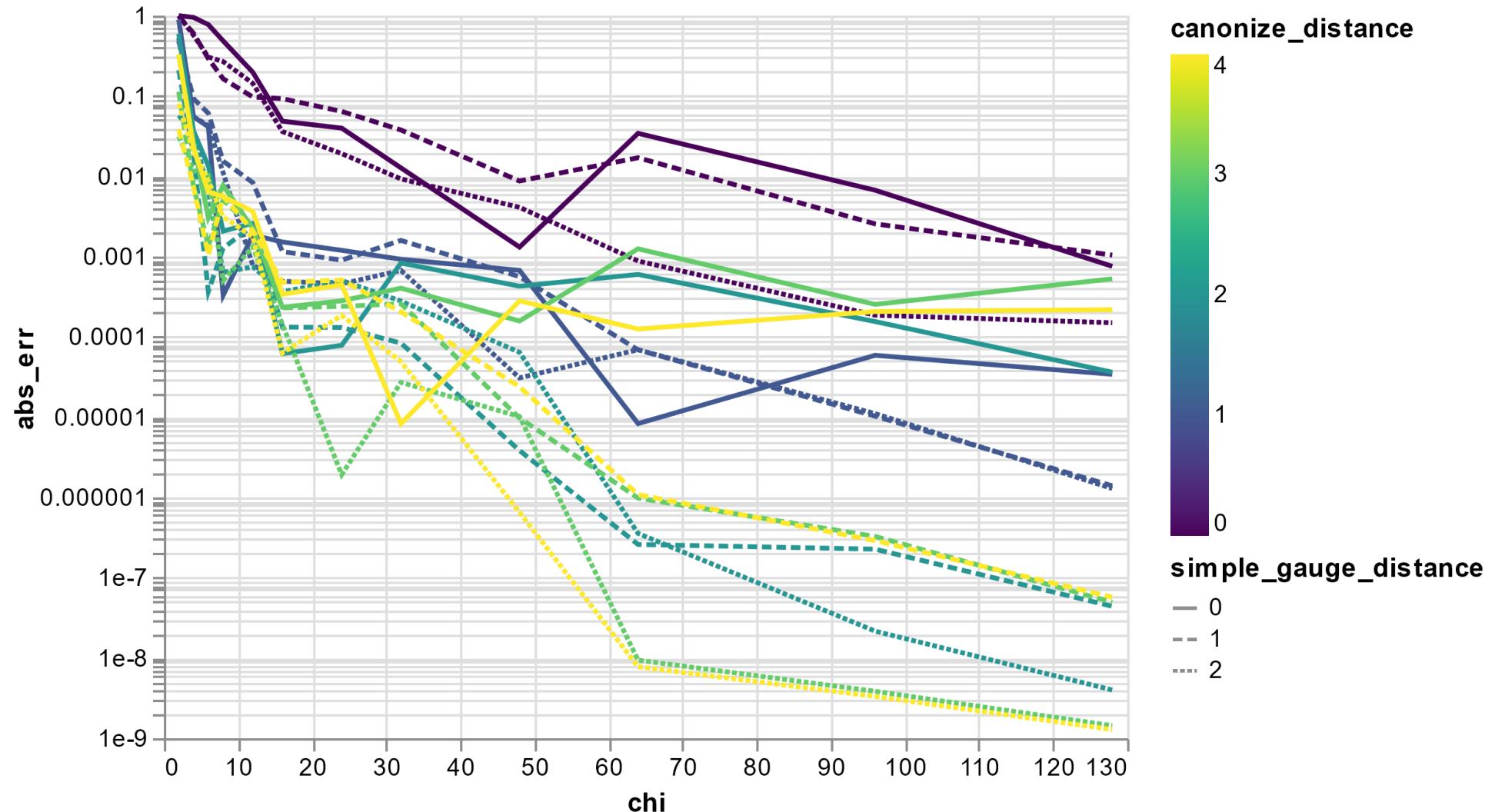
Using full-bond compression very very good

Full complex eigendecomposition seems to be required for highest accuracy

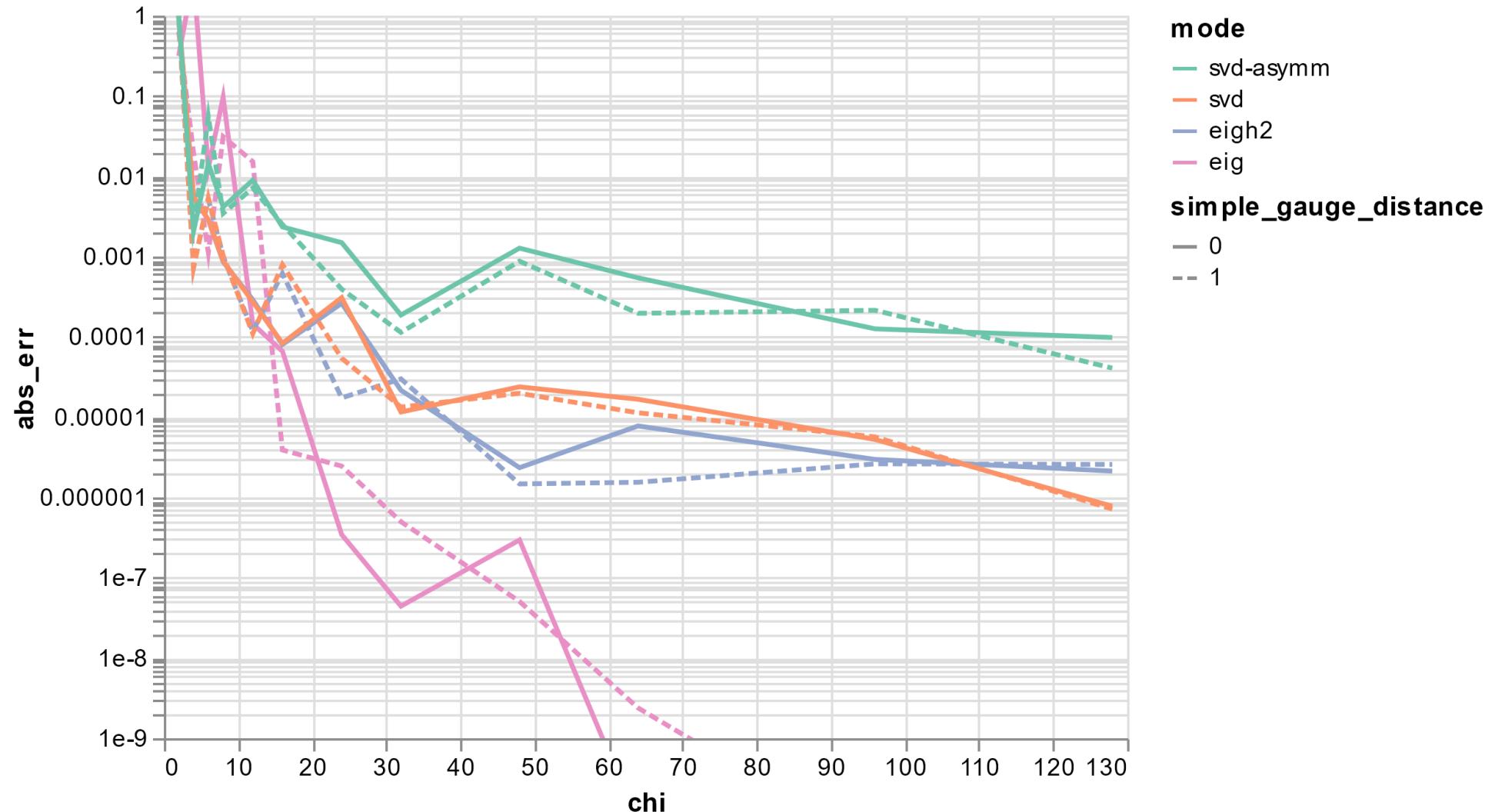
Results – 10x10 D=2 PEPS sandwich- [medium]



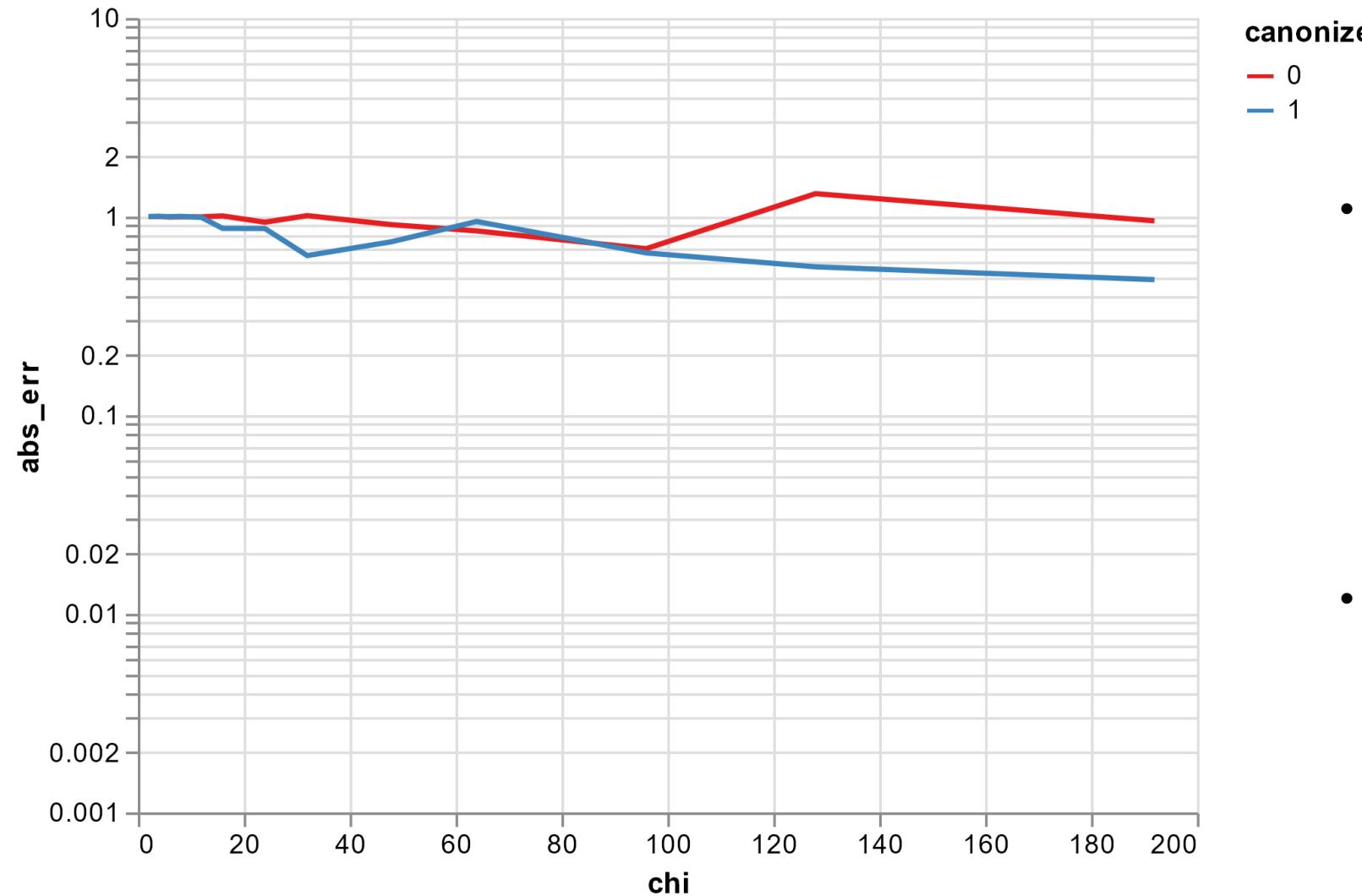
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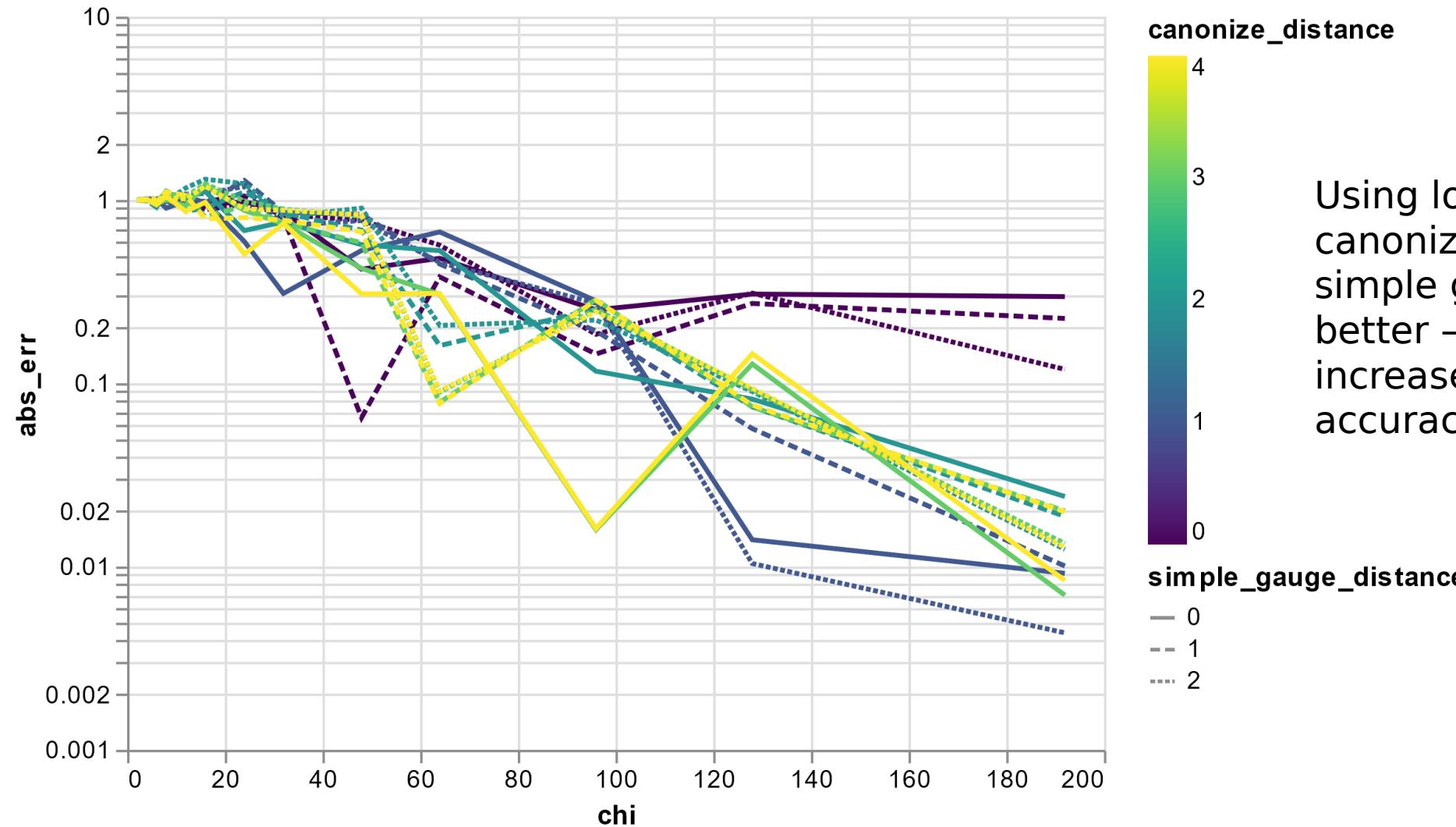


Results - 10x10 D=4 rand TN - [hard]



- Completely random 2D TN (i.e. not formed from random PEPS sandwich), much much harder
- Boundary contract not meaningful even with

Results - 10x10 D=4 rand TN - [hard]



Using local canonization and simple gauging a bit better – reliable increases for accuracy with

Results - 10x10 D=4 rand TN - [hard]

