

# Tensor network representations from the geometry of entangled states

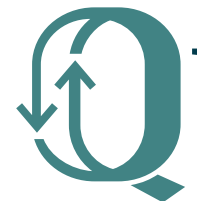
Albert H. Werner (Copenhagen)

Matthias Christandl (Copenhagen & MIT)

Angelo Lucia (Copenhagen -> Caltech)

Peter Vrana (Budapest & Copenhagen)

arXiv:1809.08185

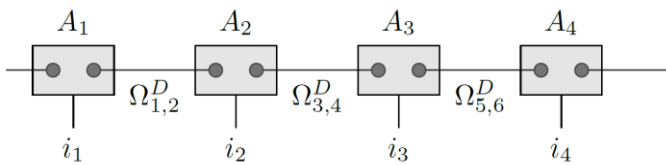


Alexander von Humboldt  
Stiftung/Foundation

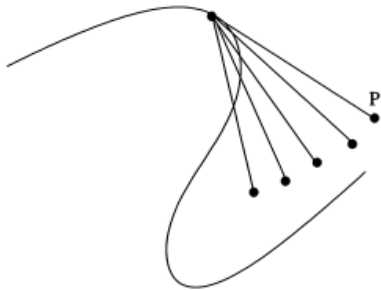
BIRS workshop: Quantum Walks and Information Tasks

# Outline

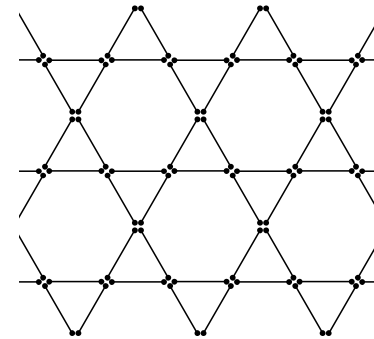
## 1. Matrix product states



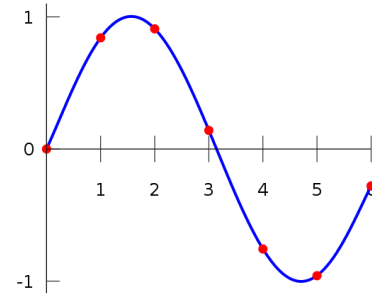
## 2. Geometry of entanglement



## 3. Tensor Networks

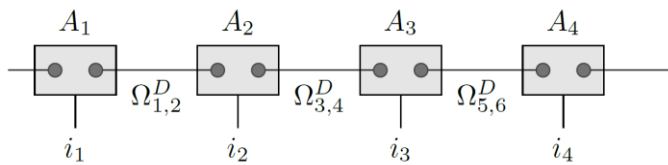


## 4. Reducing the bond dimension

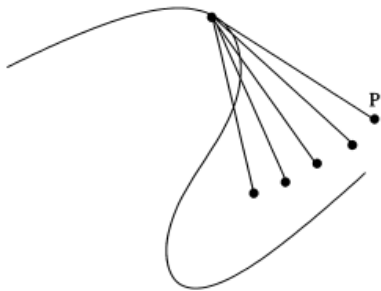


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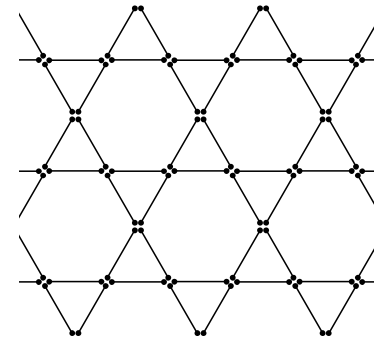
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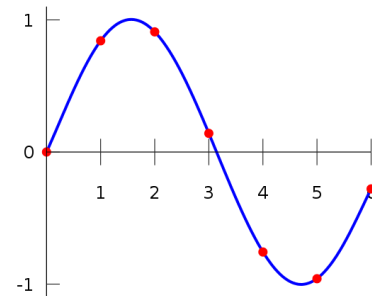
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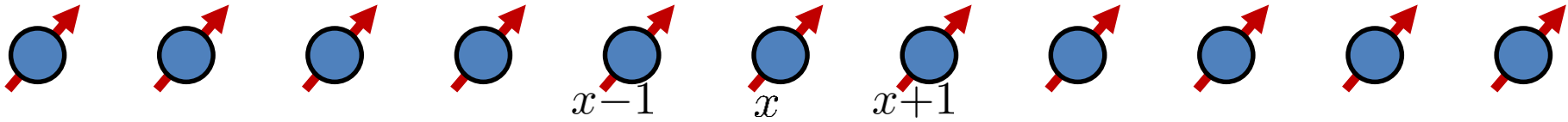
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## 4. Reducing the bond dimension

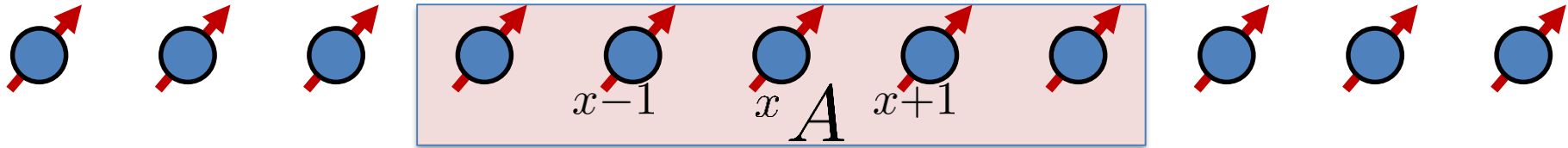


# Quantum many-body systems



- state of  $L$  spins/qudits:  $\phi \in \mathcal{H} = (\mathbb{C}^d)^{\otimes L}$ 
  - $\dim(\mathcal{H}) = d^L$
  - describes physically possible
  - what about physically reasonable?

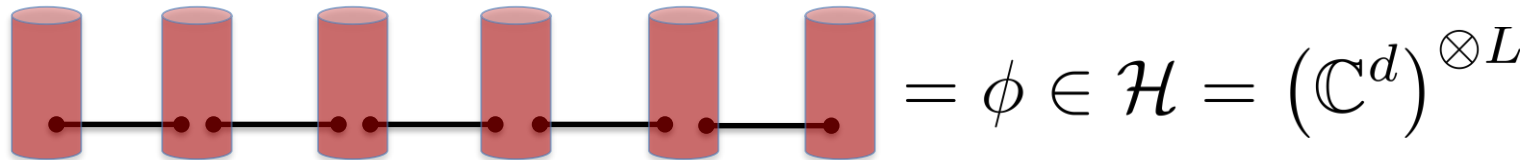
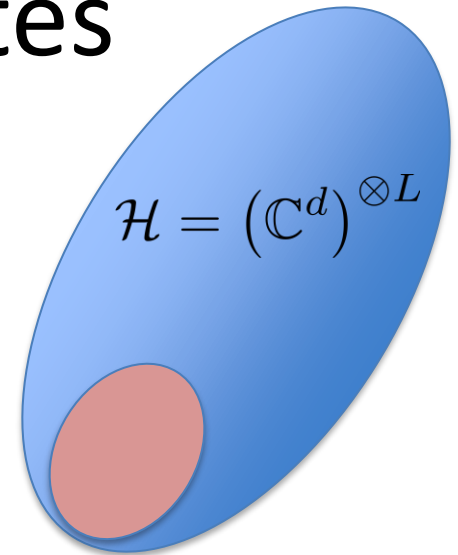
# Quantum many-body systems




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  - $\dim(\mathcal{H}) = d^L$
  - describes physically possible
  - what about physically reasonable?
- Entanglement entropy:  $S(A) = \text{tr}(\rho_A \log(\rho_A))$ 
  - random state:  $S(A) \sim \text{Vol}(A)$
  - ground states of local Hamiltonians  $S(A) \sim \text{Area}(A)$
  - area vs. volume law

# Matrix product states

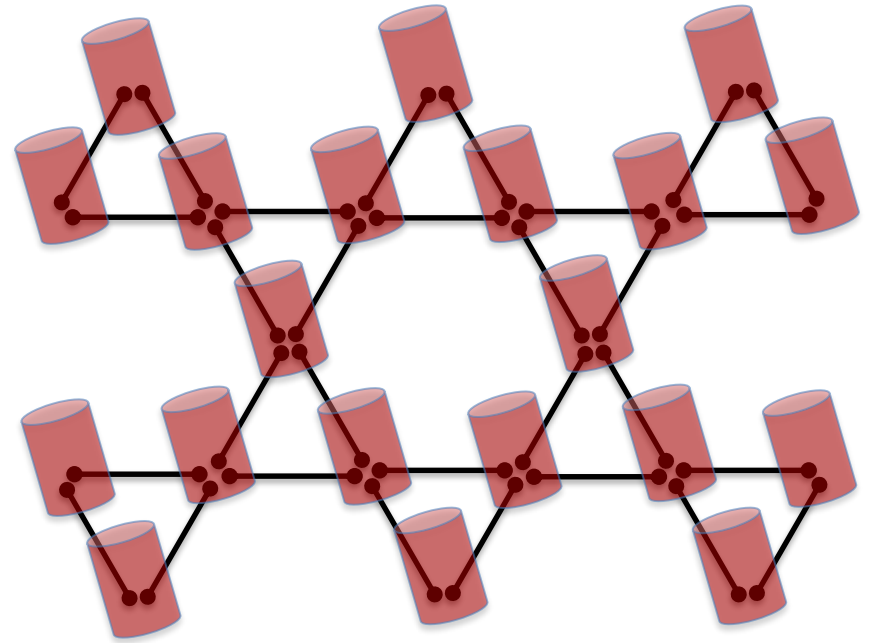
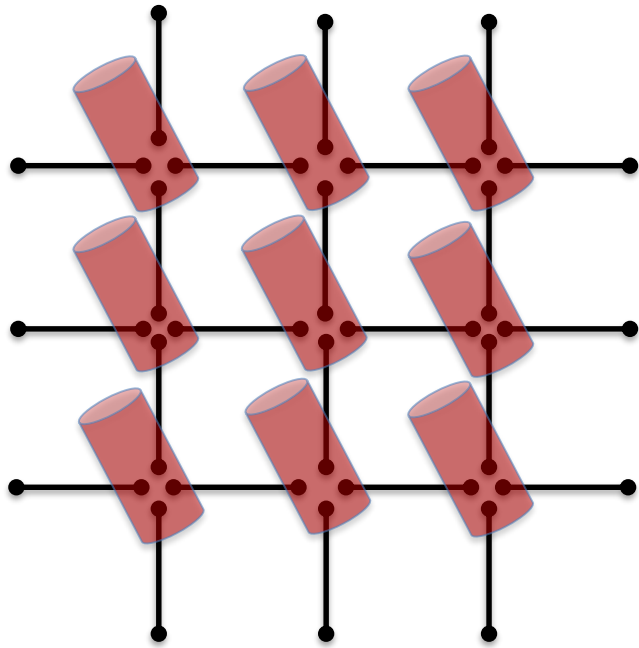
- physical corner of Hilbert space
  - find efficient parametrization
  - build in area law
- Matrix product states (MPS)



- network of max. entangled states  $\Omega^D = \sum_{l=1}^D |l, l\rangle = \bullet \text{---} \bullet$
- apply local maps:  :  $\mathbb{C}^D \otimes \mathbb{C}^D \mapsto \mathbb{C}^d$
- # of parameters  $\sim dD^2L$
- efficient approximation of groundstates

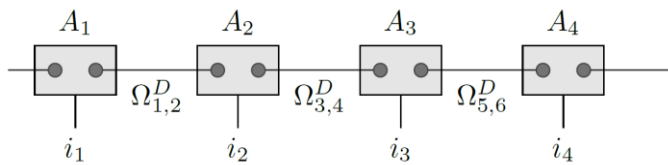
# Higher dimensions

- projected entangled pair states (PEPS)

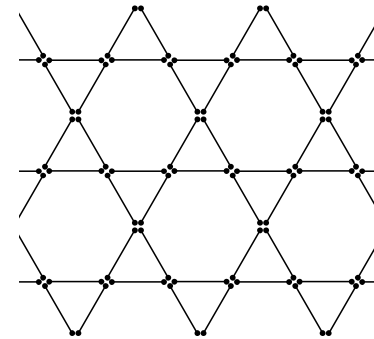


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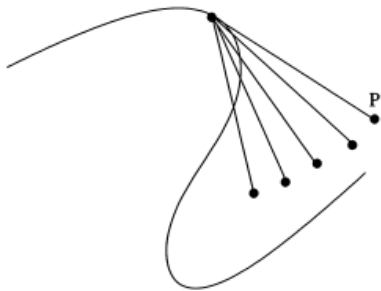
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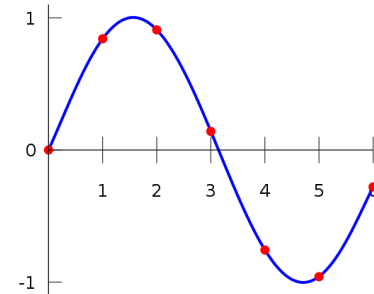
## 3. Tensor Networks



## 2. Geometry of entanglement



## 4. Reducing the bond dimension

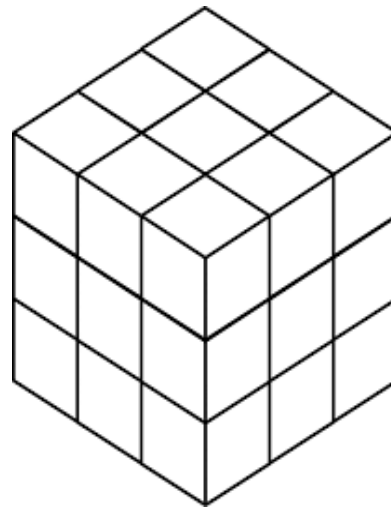
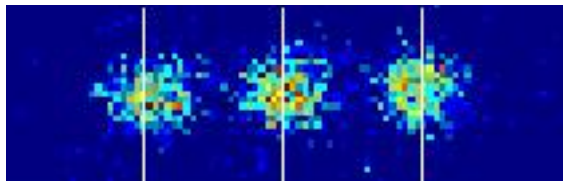




# Quantum state=tensor

$$t \in \mathbb{C}^d \otimes \mathbb{C}^d \otimes \mathbb{C}^d$$

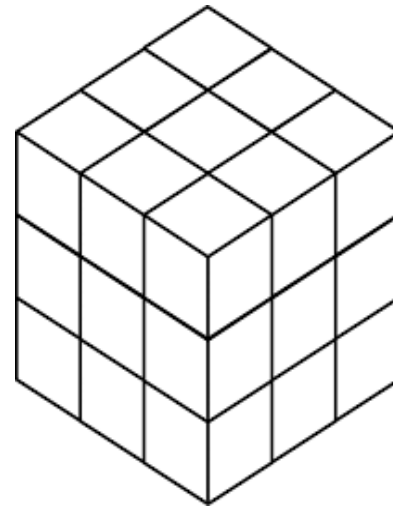
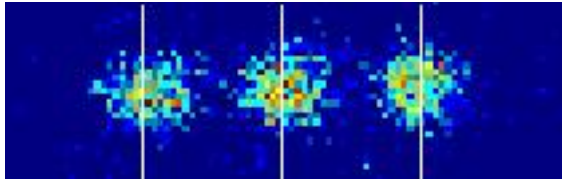
$$t = \sum_{i,j,k=1}^d t_{ijk} e_i \otimes e_j \otimes e_k$$



# Local operations=restrictions

$$t \geq t' \text{ if } (a \otimes b \otimes c) t = t'$$

for some matrices  $a, b, c$



Linear combination of slices

# Restriction

$$t \geq t' \text{ if } (a \otimes b \otimes c) t = t'$$

for some matrices  $a, b, c$

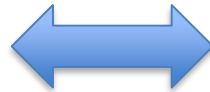
$$t \cong t' \text{ if } t \geq t' \text{ and } t' \geq t$$

$$\text{iff } (a \otimes b \otimes c) t = t'$$

for invertible  $a, b, c$

$$\text{iff } G.t = G.t' \quad \leftarrow G = GL(d) \times GL(d) \times GL(d)$$

Deciding restriction



Classifying orbits  
and their relations

# 3 qubits

Greenberger-Horne-Zeilinger  
GHZ-state

Einstein-Podolsky-Rosen  
(EPR)-state

$$e_0 \otimes e_0 \otimes e_0 + e_1 \otimes e_1 \otimes e_1$$

W-state

$$\approx e_0 \otimes e_0 \otimes e_1 + e_0 \otimes e_1 \otimes e_0 + e_1 \otimes e_0 \otimes e_0$$

$$e_0 \otimes e_0 \otimes e_0 + e_1 \otimes e_1 \otimes e_0$$

$$e_0 \otimes e_0 \otimes e_0 + e_0 \otimes e_1 \otimes e_1$$

$$e_0 \otimes e_0 \otimes e_0 + e_1 \otimes e_0 \otimes e_1$$

$$e_0 \otimes e_0 \otimes e_0$$

unentangled state

GHZ state



# Degeneration

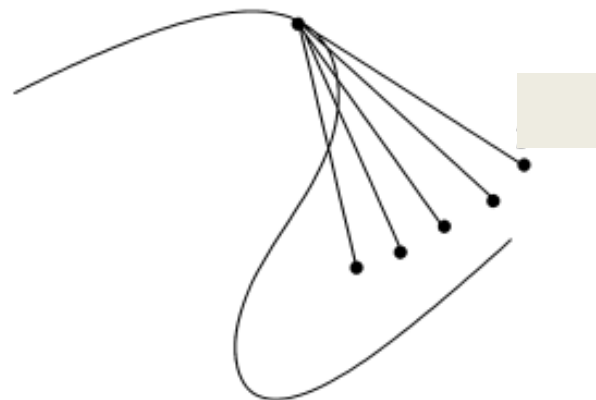
$$(e_0 + \epsilon e_1)^{\otimes 3} - e_0^{\otimes 3}$$

$$= \epsilon(e_0 \otimes e_0 \otimes e_1 + e_0 \otimes e_1 \otimes e_0 + e_1 \otimes e_0 \otimes e_0) + O(\epsilon^2)$$

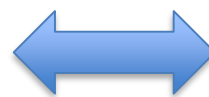
W state



$$t \trianglelefteq t' \text{ if } t_\epsilon \xrightarrow[\epsilon \mapsto 0]{} t', t \geq t_\epsilon$$



Deciding degeneration



Classifying orbit closures and their relations

# Deciding degeneration

- Orbit closures are  $G$ -invariant algebraic varieties

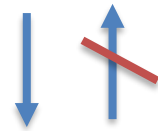
$t \not\preceq t'$  iff there exists

$G$  – covariant polynomial  $f$  :

$$f(t) = 0, \text{ but } f(t') \neq 0$$

- **Example:**

$$e_0 \otimes e_0 \otimes e_0 + e_1 \otimes e_1 \otimes e_1$$



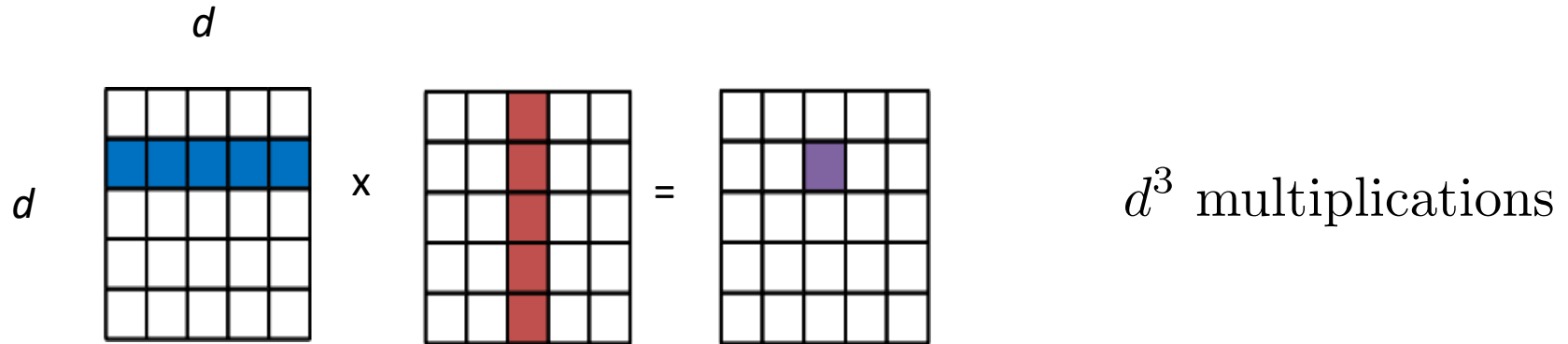
$f = \text{Cayley hyperdeterminant}$

$$\approx e_0 \otimes e_0 \otimes e_1 + e_0 \otimes e_1 \otimes e_0 + e_1 \otimes e_0 \otimes e_0$$

# Algebraic Complexity

$M(d) =$  algebra of  $d \times d$  complex matrices

$Mamu(d) : M(d) \times M(d) \rightarrow M(d)$       bilinear  
 $(A, B) \mapsto A \cdot B$



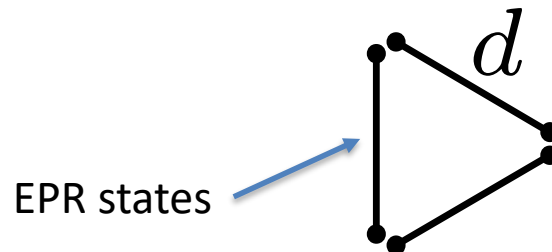
# Bilinear maps=tensors

$$\text{Mamu}(d) : M(d) \times M(d) \times M(d)^* \rightarrow \mathbf{C}$$

$$(A, B, C) \mapsto \text{tr} A \cdot B \cdot C$$

$$\text{Mamu}(d) = \sum_{i,j,k=1}^d e_{ij} \otimes e_{jk} \otimes e_{ki} \quad \leftarrow e_{ij} = e_i \otimes e_j$$

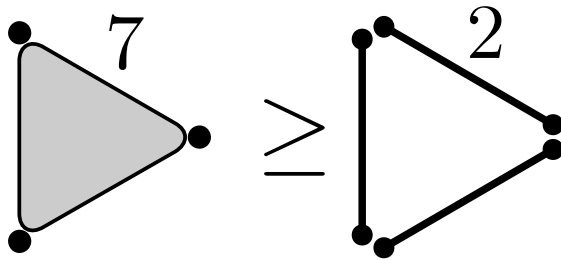
$$= \sum_{i,j,k=1}^d (e_i \otimes e_j) \otimes (e_j \otimes e_k) \otimes (e_k \otimes e_i)$$





# Complexity=Tensor rank

Strassen: # elementary multiplications = tensor rank



In this context, many techniques have been developed to understand restriction and degeneration

$$e_{00} \otimes e_{00} \otimes e_{00} + e_{11} \otimes e_{11} \otimes e_{11}$$

$$e_{01} \otimes e_{10} \otimes e_{00} + e_{10} \otimes e_{01} \otimes e_{11}$$

$$e_{01} \otimes e_{11} \otimes e_{10} + e_{10} \otimes e_{00} \otimes e_{01}$$

$$e_{00} \otimes e_{01} \otimes e_{10} + e_{11} \otimes e_{10} \otimes e_{01}$$

$$e_{\pm} := e_0 \pm e_1$$

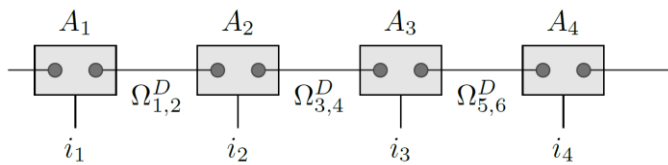
$$= e_{-1} \otimes e_{1+} \otimes e_{00} + e_{1+} \otimes e_{00} \otimes e_{-1} + e_{00} \otimes e_{-1} \otimes e_{1+}$$

$$- e_{-0} \otimes e_{0+} \otimes e_{11} - e_{0+} \otimes e_{11} \otimes e_{-0} - e_{11} \otimes e_{-0} \otimes e_{0+}$$

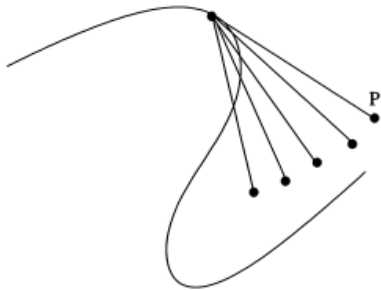
$$+ (e_{00} + e_{11}) \otimes (e_{00} + e_{11}) \otimes (e_{00} + e_{11})$$

# Outline

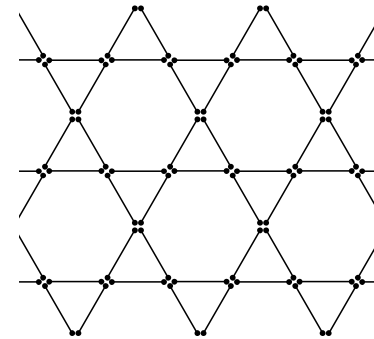
## 1. Matrix product states



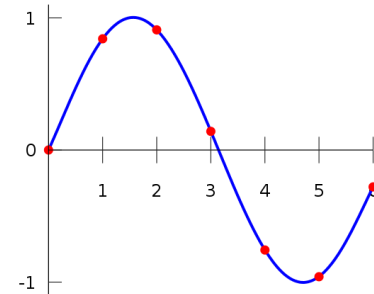
## 2. Geometry of entanglement



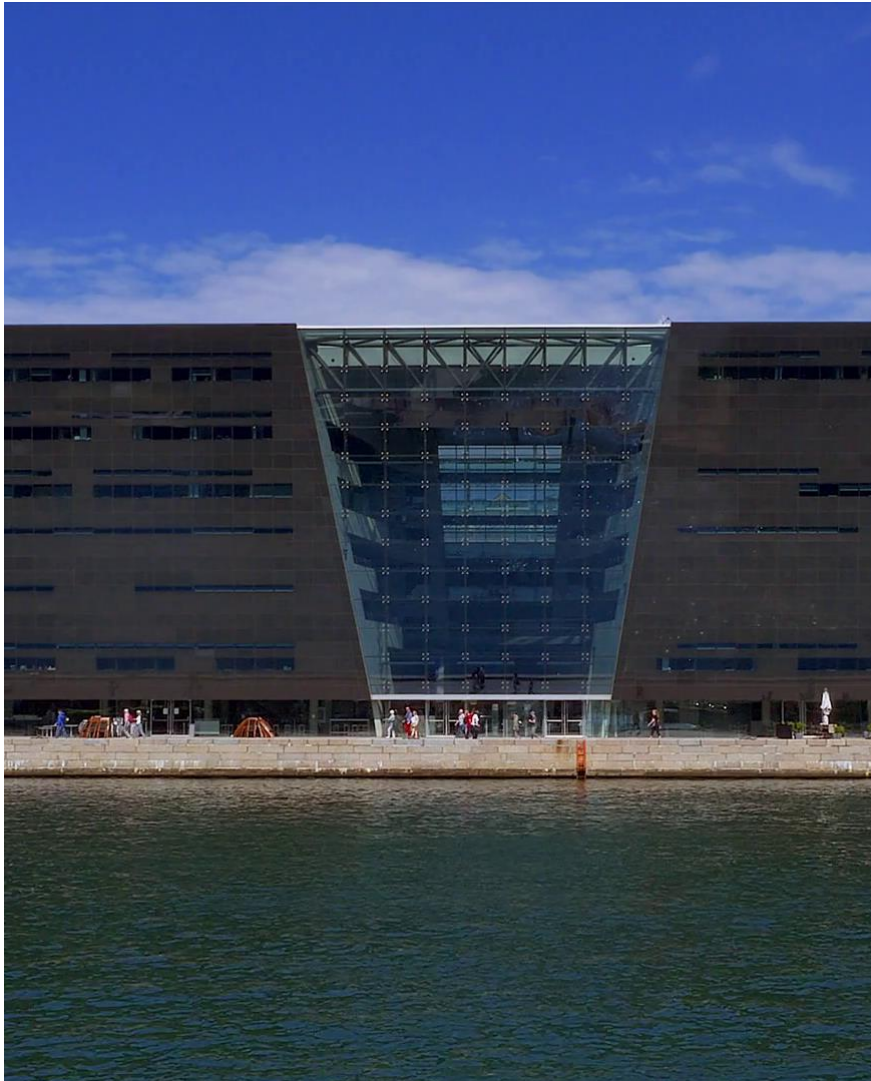
## 3. Tensor Networks



## 4. Reducing the bond dimension



# Graph or Hypergraph

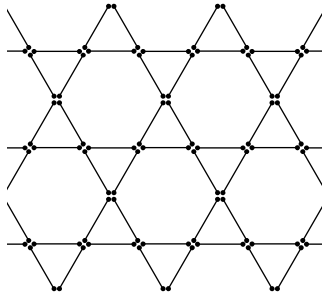


# Tensor networks

line or circle=MPS  
lattice=PEPS

lattices: e.g. Chen et al.'11, Xie et al.'14  
Schuch et al.'12, Molnar et al.'18

- Choose graph



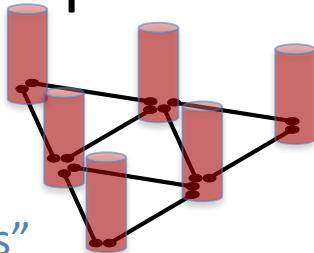
- Associate entangled state to edges

$$\Omega^D = \sum_{l=1}^D |l, l\rangle$$

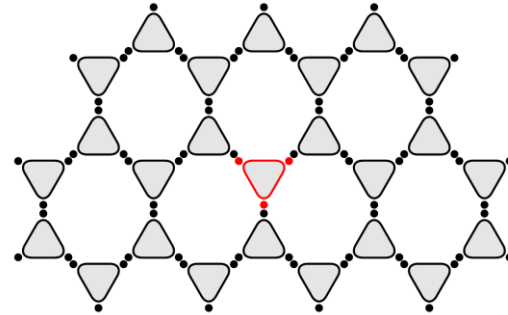
← bond dimension

- Apply linear maps to vertices

tensors in  
"tensor networks"



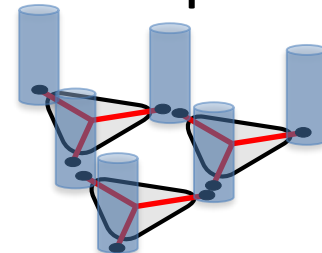
- Choose hypergraph



- Associate entangled state to hyperedges

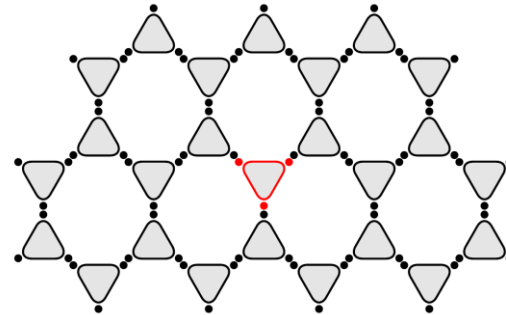
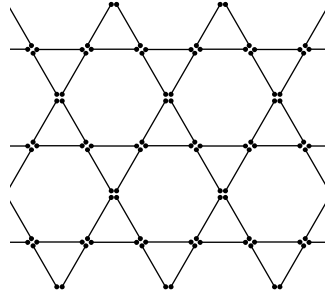
$$\sum_{i=1}^k |i\rangle |i\rangle |i\rangle = \text{triangle}(k) \cdot \sum_{i,j,k=0}^2 \varepsilon_{i,j,k} |i, j, k\rangle + |2, 2, 2\rangle = \text{triangle}(3)$$

- Apply linear maps to vertices



# Tensor networks

- Underlying “entanglement structure”



- Does the job

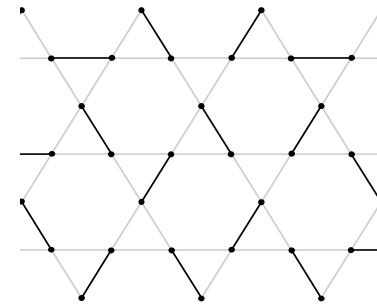
- E.g. represents Resonating Valence Bond state

- Verstraete et al.'06 

- Schuch et al.'12 

- You don't like it

- Too large bond dimension
- Weird entangled state



Contraction is too slow  
Code does not work at all  
(you are computational physicist)

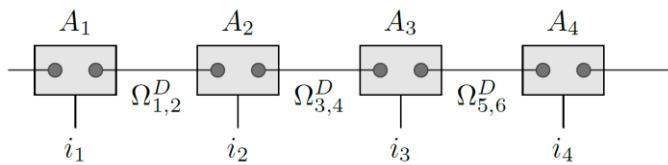
Representation is too ugly  
(you are mathematical physicist)

# Let's find a better tensor network!

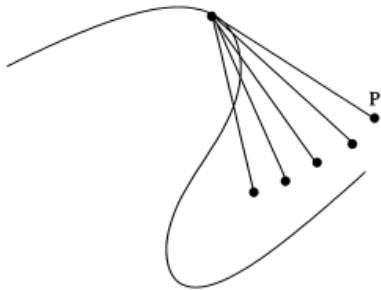
- Start from scratch
  - I.e. from physical state
  - Pro: super-optimized
  - Con: are you kidding?
- Switch entanglement structure
  - Independent from projectors
  - Pro: Works for any state with same structure
  - Pro: Tight for injective ones
  - Con: Tailored optimization could be better

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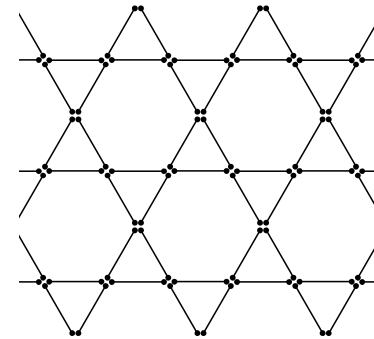
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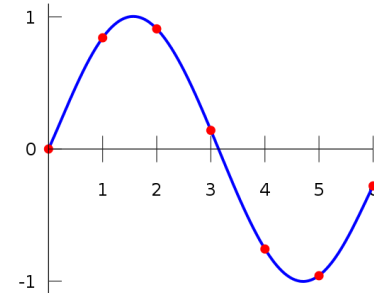
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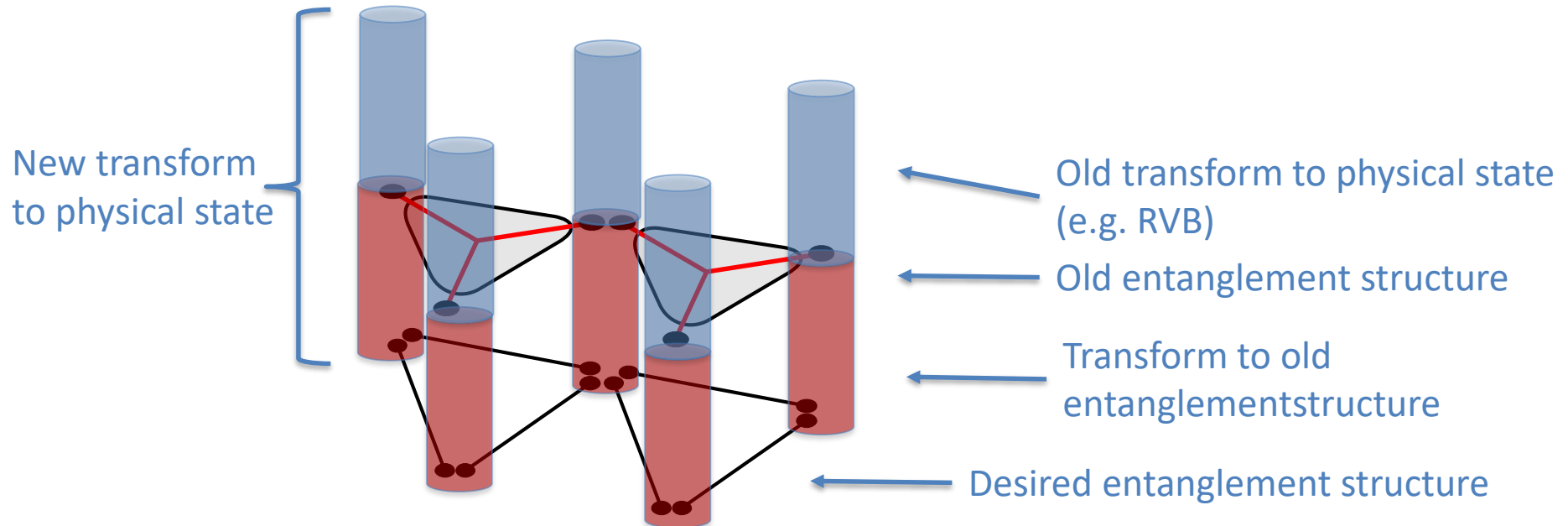
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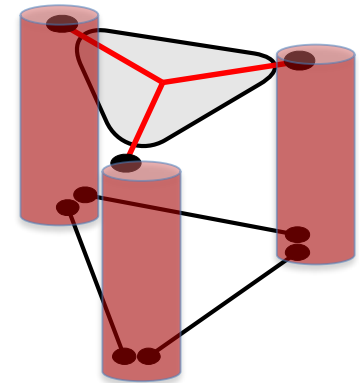
## 4. Reducing the bond dimension



# Switch entanglement structure



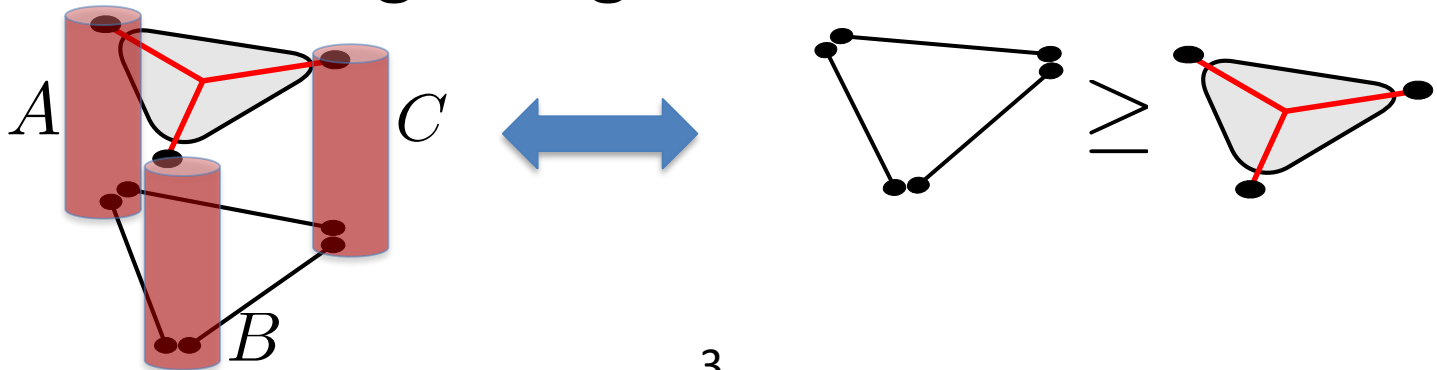
- Let us just focus on transforming the entanglement structure
- Plaquette by plaquette



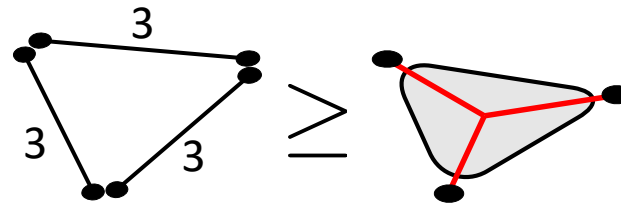


# Entanglement

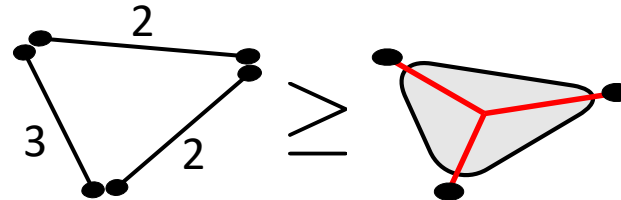
- Back to the beginning



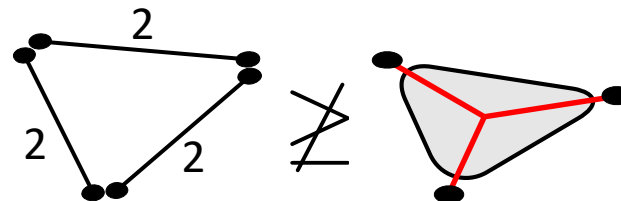
- State of the art
  - Molnar et al.'18



- Improvement

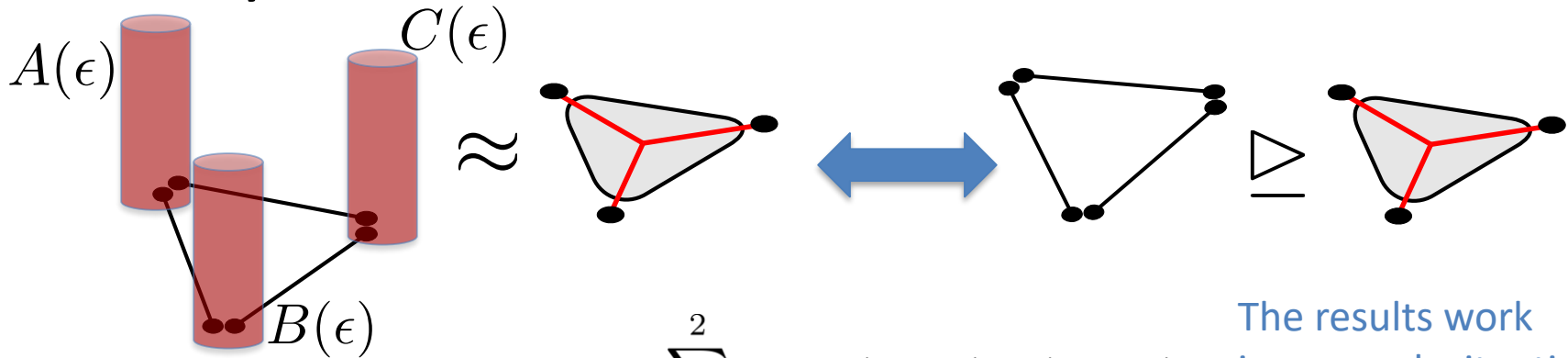


- But not all the way

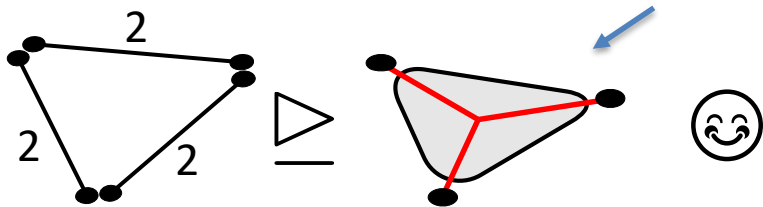


# Entanglement

- Why not relax?



- It works!



$$\sum_{i,j,k=0}^2 \varepsilon_{i,j,k} |i, j, k\rangle + |2, 2, 2\rangle$$

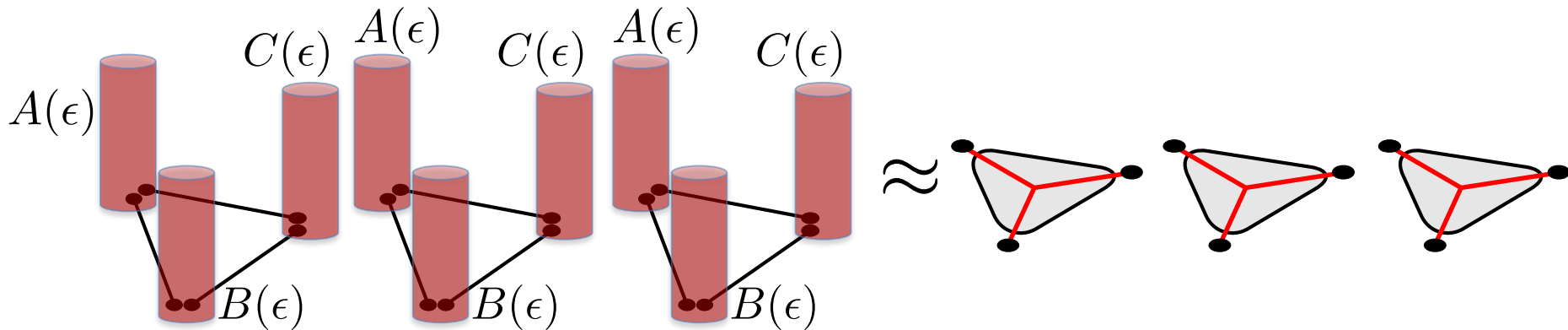
The results work in any such situation

- We are looking for approximate MPS
- Pauli's give exact MPS for antisymm
- Squeeze in an epsilon

$$\varepsilon^2 \text{triangle} + \varepsilon^4 |2\rangle \otimes \left( \frac{1}{4} |00\rangle - |11\rangle \right)$$

$$M_0^{[j]}(\varepsilon) = \frac{1}{2} \begin{pmatrix} 0 & \varepsilon \\ \varepsilon & 0 \end{pmatrix}, \quad M_1^{[j]}(\varepsilon) = \begin{pmatrix} 0 & -\varepsilon \\ \varepsilon & 0 \end{pmatrix}, \quad M_2^{[j]}(\varepsilon) = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} + \frac{\delta_{j,3}\varepsilon^2}{2} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

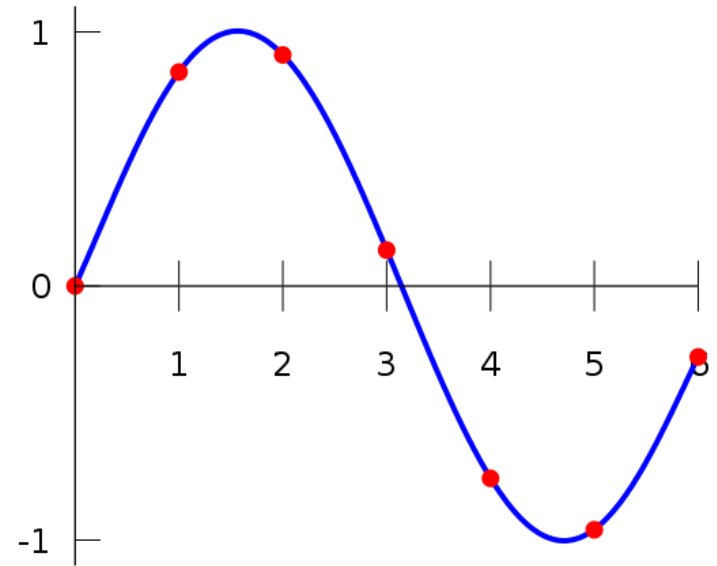
# Roll it out



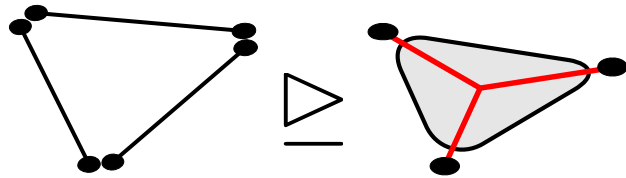
- Leads to approximate tensor network representation
- Who cares?
- Let's convert it to an exact one!
  - Only need to pay a small price

# Interpolation

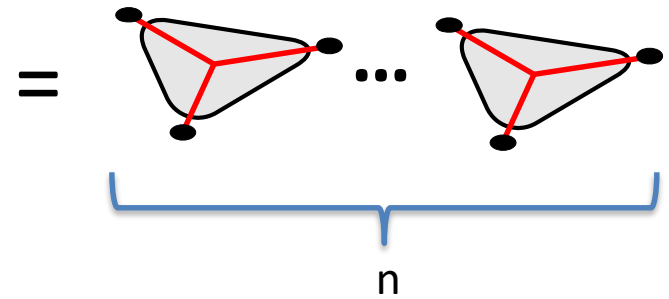
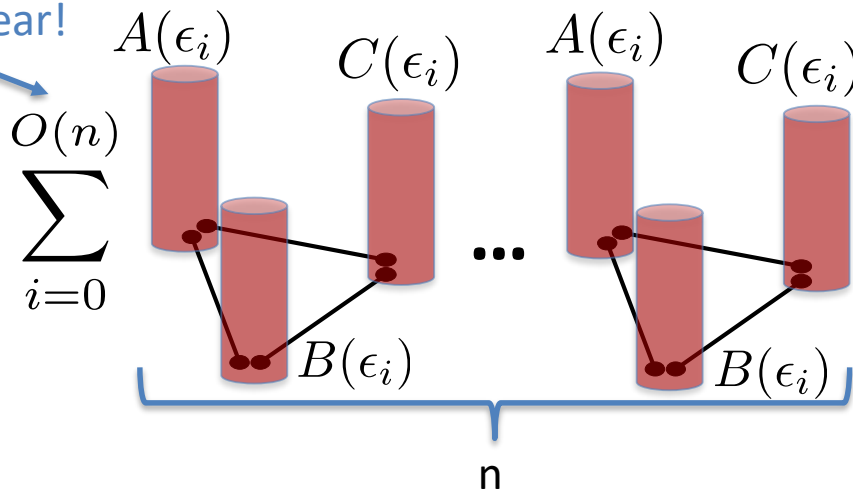
- Given a polynomial  $p(x)$  of degree  $n$
- Obtain  $p(0)$  by Lagrange interpolation
  - Evaluate at  $n+1$  points
  - Determines the entire polynomial
  - Value at 0 can be easily obtained (see Wiki)
- Our transformation matrices are polynomial in epsilon!



# Lemma: Degeneration $\rightarrow$ Restriction



Only linear!

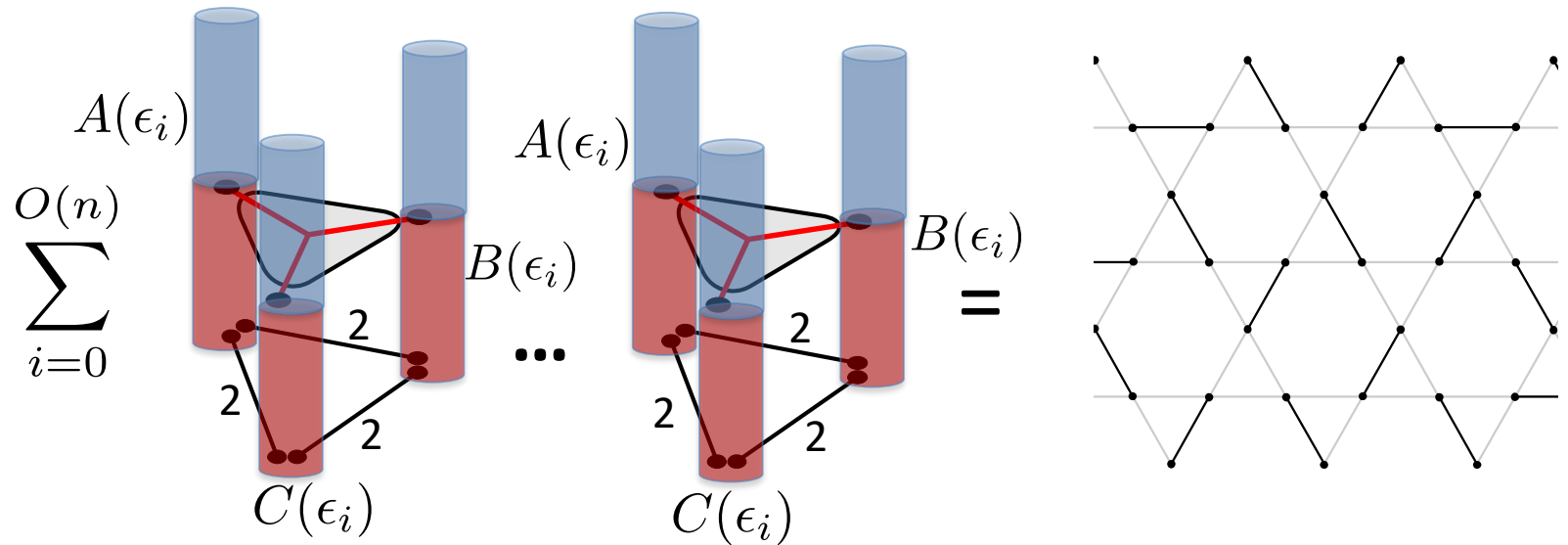


- Proof based on interpolation
  - Bini, Lotti and Romani, SIAM J.Comp. 1980
  - Christandl, Jensen & Zuiddam, Lin.Alg. App 2018

It's about tensor rank

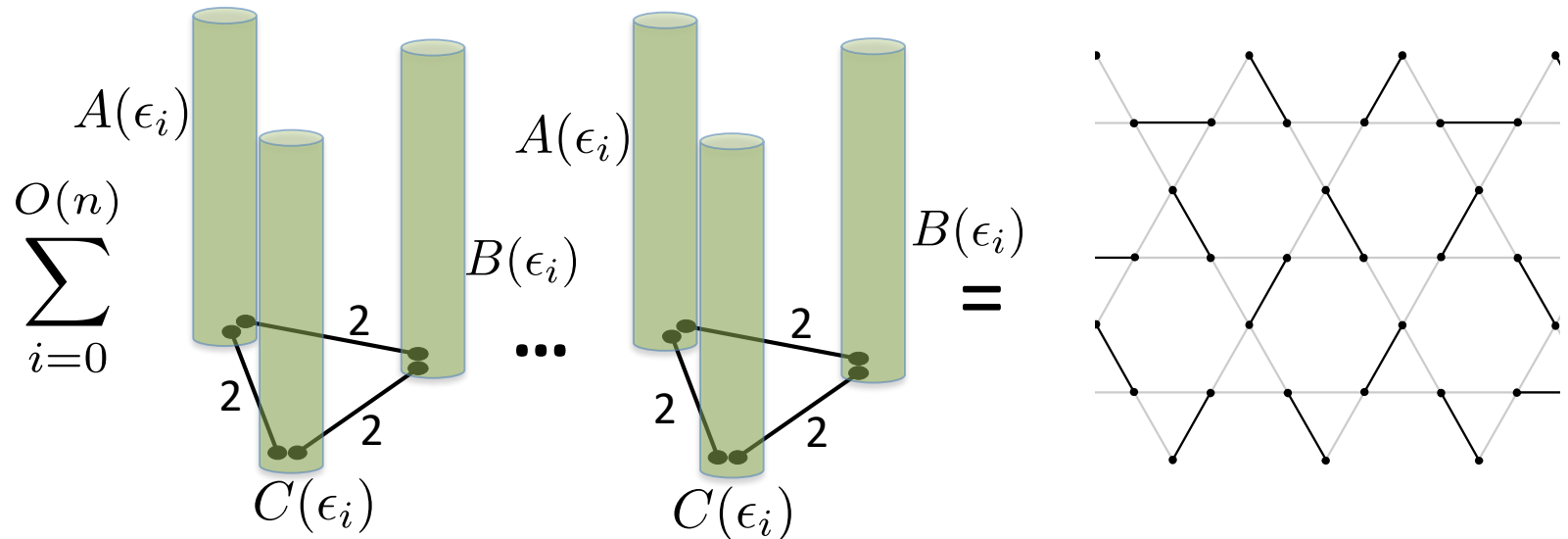
# Application

- Resonating Valence Bond State



# Application

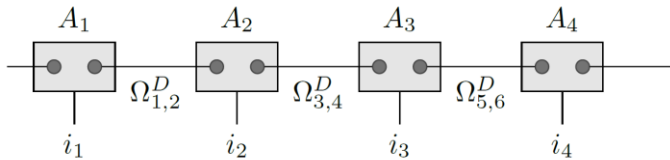
- Resonating Valence Bond State



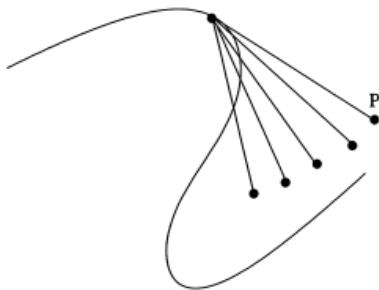
- Parallel algorithm for faster contraction
  - also for expectation values

# Summary

## 1. Matrix product states

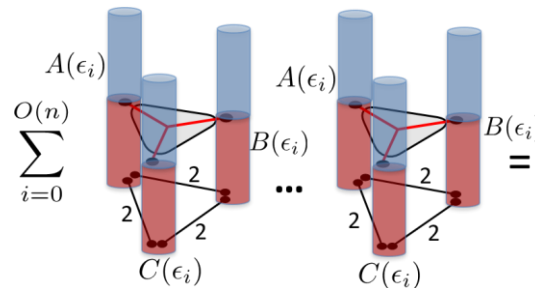


## 2. Geometry of entanglement

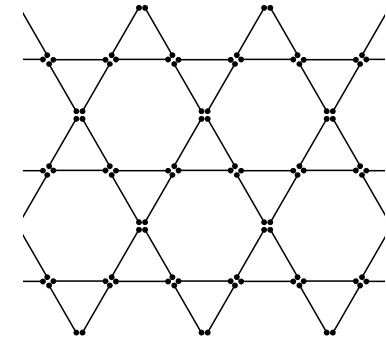


## Open questions

- In practice?
- Other examples?
- Sums of tensor networks as new variational class?



## 3. Tensor Networks



## 4. Reducing the bond dimension

