On the tensor rank of networks of entangled pairs

tensor surgery and the laser method



Peter Vrana (Budapest) and Jeroen Zuiddam (Amsterdam) arXiv:1606.04085 and arXiv:1609.07476



Prelude

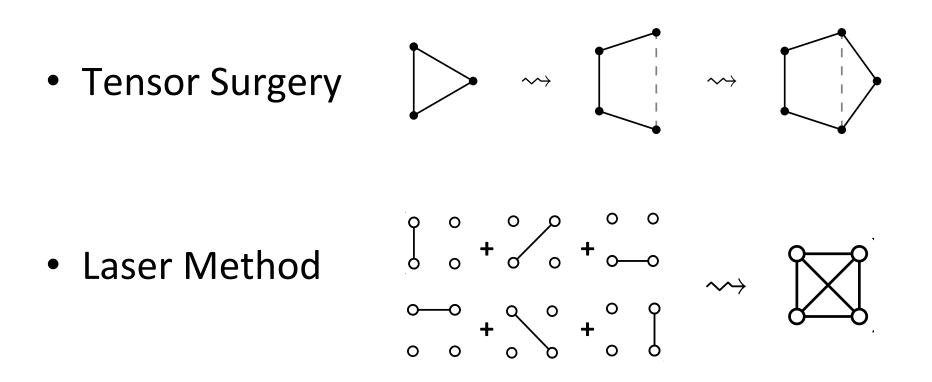
- Quantum Information $|000\rangle + |111\rangle$ Tensors = multiparticle quantum states Tensor rank = multiparticle entanglement
- Computer Science Communication complexity Tensors = description of algebraic problems

 Tensor rank = complexity
- Physics
 A1 A2 A3 A4 A5
 Wiki Tensor networks = description of ground states
 Tensor structure = physical properties
- Relativity Theory, Engineering, ...



Overview

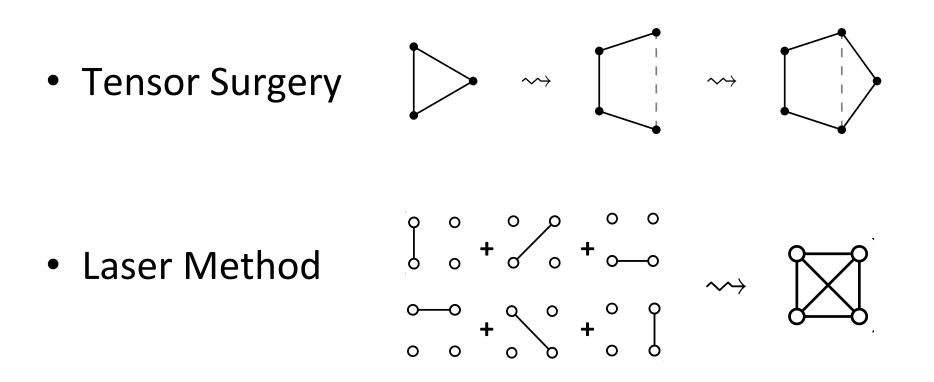
• Tensors, rank and networks of entangled pairs





Overview

• Tensors, rank and networks of entangled pairs





Tensors

tensor=quantum state overall normalisation

- k-tensor t is an element in $\mathbb{C}^{n_1} \otimes \cdots \otimes \mathbb{C}^{n_k}$ basis $b_{i_1} \otimes \cdots \otimes b_{i_n}$ k=1: vector $|+\rangle = |0\rangle + |1\rangle = b_0 + b_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ $b_i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$ • **k=2: matrix** $|00\rangle + |11\rangle = b_0 \otimes b_0 + b_1 \otimes b_1 = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ $b_{i_1} \otimes b_{i_2} = \left(\begin{array}{ccc} 0 & \dots & 0\\ \vdots & \ddots & 1\\ 0 & \dots & 0\end{array}\right)$
- k=3: cube $|001\rangle + |010\rangle + |100\rangle = b_0 \otimes b_0 \otimes b_1 + b_0 \otimes b_1 \otimes b_0 + b_1 \otimes b_0 \otimes b_0$

$$b b_0 + b_1 \otimes b_0 \otimes b_0$$

 $b_{i_1} \otimes b_{i_2} \otimes b_{i_3} =$



Resource theory of tensors

(Strassen 1991, Dür, Vidal & Cirac 2000)

SLOCC

- Given k-tensors s and t.
- Can s be transformed into t by local operations? Do matrices A_1, \ldots, A_k exist, so that

$$A_1\otimes\cdots A_ks=t$$
 ?

- If yes, we write $\ s\geq t$.
- unit tensor (GHZ entangled state) $T_d(k) = \sum_{i=1}^d b_i \otimes \cdots \otimes b_i$



Resource theory of tensors

- "entanglement cost" of a k-tensor t is the smallest d, s.th. $T_d(k) \geq t$
- equals tensor rank R(t) $R(t) = \min\{d : t = \sum_{i} v_1^{(i)} \otimes \cdots \otimes v_k^{(i)}\}$
- "distillable entanglement" of a k-tensor t is the largest d, s.th. $t \ge T_d(k)$

equals subrank



Tensor rank

- k=2: tensor rank=matrix rank
 → easy to compute
 Example:
 Identity matrix (EPR state) T(-) = ∑ b_i ⊗ b_i ∈ C² ⊗ C²
- k>2: NP hard (Håstad) Examples: Rank 2: GHZ state $T(\mathbf{b} \cdot) = \sum_{i \in \{0,1\}} b_i \otimes b_i \otimes b_i \in \mathbb{C}^2 \otimes \mathbb{C}^2 \otimes \mathbb{C}^2$

Rank 3: W-state $b_0 \otimes b_0 \otimes b_1 + b_0 \otimes b_1 \otimes b_0 + b_1 \otimes b_0 \otimes b_0$

Tensor rank versus tensor product

- Given s and t both k-tensors $R(s \otimes t) \leq R(s)R(t)$
- Equality for 2-tensors, strict in general

$$T\left(\begin{array}{c} & \\ \bullet \end{array}\right) = \sum_{i \in \{1,2\}} b_i \otimes b_i \otimes 1 \in \mathbb{C}^2 \otimes \mathbb{C}^2 \otimes \mathbb{C},$$
$$T\left(\begin{array}{c} & \\ \bullet \end{array}\right) = \sum_{i \in \{1,2\}} b_i \otimes 1 \otimes b_i \in \mathbb{C}^2 \otimes \mathbb{C} \otimes \mathbb{C}^2,$$
$$T\left(\begin{array}{c} & \\ \bullet \end{array}\right) = \sum_{i \in \{1,2\}} 1 \otimes b_i \otimes b_i \in \mathbb{C} \otimes \mathbb{C}^2 \otimes \mathbb{C}^2.$$

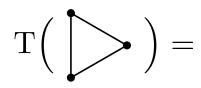
2x2 matrix multiplication tensor, rank =7 (Strassen)

$$\mathrm{T}\Big(\bigcup \Big) = \sum_{i,j,k \in \{1,2\}} (b_i \otimes b_j) \otimes (b_j \otimes b_k) \otimes (b_k \otimes b_i)$$



Strassen's 7er

• Define $b_+ \coloneqq b_0 + b_1$ and $b_- \coloneqq b_0 - b_1$ in \mathbb{C}^2 $b_{xy} \coloneqq b_x \otimes b_y \in \mathbb{C}^2 \otimes \mathbb{C}^2$



takes half an hour to verify



Border rank

 Sometimes, tensor of rank r can be approximated arbitrarily by tensors of rank b<r

$$(b_0 + \epsilon b_1)^{\otimes 3} - b_0^{\otimes 3}$$

= $\epsilon (b_0 \otimes b_0 \otimes b_1 + b_0 \otimes b_1 \otimes b_0 + b_1 \otimes b_0 \otimes b_0) + O(\epsilon^2)$

- Smallest b is called border rank $\underline{R}(t)$
- More generally, approximate transformation from s to t.



Resource theory of tensors

- Given k-tensors s and t. When is $s^{\otimes m} \ge t^{\otimes n}$?
- Best ratio m/n denoted by $\omega(s,t)$
- Asymptotic log rank $\omega(t) := \omega(T_2(k), t)$
- Asymptotic log subrank $q(t) := \omega(t, T_2(k))^{-1}$
- Theorem (Strassen & co):

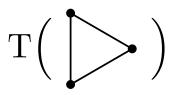
 $q(t) \le \omega(t) \le \log \underline{R}(t) \le \log R(t)$



 $q(t) \le \omega(t) \le \log \underline{R}(t) \le \log R(t)$

matrix multiplication exponent

• Mamu



$$q(t) = 2$$

Strassen

Coppersmith-Winograd, ... Le Gall

 $\omega(t) \le 2.38$

 $\underline{R}(t) = 7 \qquad R(t) = 7$ Landsberg

W state

 $b_0 \otimes b_0 \otimes b_1 + b_0 \otimes b_1 \otimes b_0 + b_1 \otimes b_0 \otimes b_0$

$$q(W) = h\left(\frac{1}{3}\right) \approx 0.92 \quad \omega(t) = 1$$

Coppersmith-Winograd

 $\underline{R}(t) = 2 \qquad R(t) = 3$



Motivation for our work

- Log rank is a lower bound on the quantum communication complexity of a function f(x,y)
- Exact for non-deterministic case Equality game = unit tensor Pairwise equality (among 3) = Mamu Savings over the classical case: $\log_2 7 < 3$
- What about more players? Pairwise equality in a circle or graph.

with Buhrman and Zuiddam, 1603.03757 Proceedings of ITCS 2017

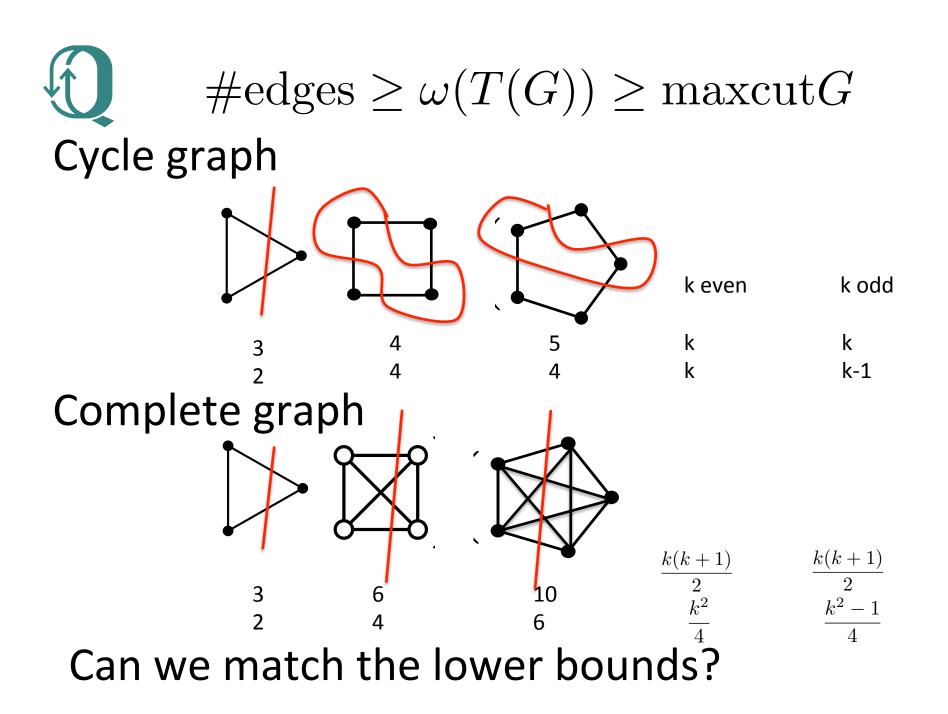


Benchmark: lower bound

- Rank, Border rank and asymptotic rank are decreasing under grouping of particles
- Group k particles into set S and complement $R(t) \geq R_S(t)$
- this is matrix rank \rightarrow easy to compute
- For a graph of entangled pairs

 $R_S(t) = 2^{\# \text{edges leaving}S} \quad \omega(T(G)) \ge \text{maxcut}G$

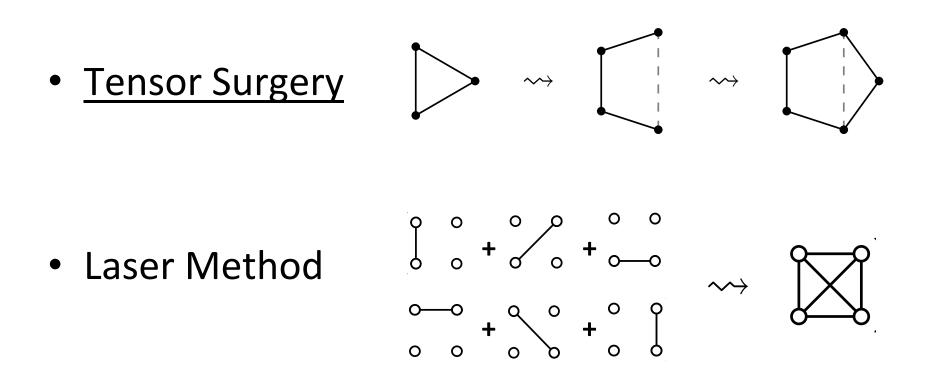
• Upper bound = # edges in graph





Overview

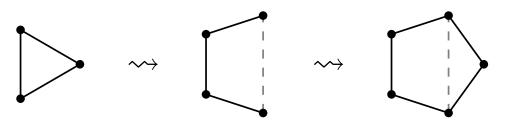
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Tensor surgery

- Transform non-trivial decomposition of initial tensor into non-trivial decomposition of target tensor
- Example: rank 7 decomposition of 3 cycle
 → rank 31 decomposition of 5 cycle
- Procedure:
 - 1) cut open, generates virtual pair
 - 2) insert 2 pairs and absorb virtual pair



$$\begin{split} & \overbrace{\mathbf{C}} & \xrightarrow{} & \overbrace{\mathbf{C}}^2 \otimes \mathbb{C}^2 \to (\mathbb{C}^2 \otimes \mathbb{C}^2)^{\otimes 3} \\ & f: \mathbb{C}^2 \otimes \mathbb{C}^2 \to (\mathbb{C}^2 \otimes \mathbb{C}^2)^{\otimes 3} \\ & I \otimes v \mapsto \sum_{j \in \{0,1\}^2} (u \otimes b_{j_1}) \otimes (b_{j_1} \otimes b_{j_2}) \otimes (b_{j_2} \otimes v). \\ & = -\phi(b_{-0}) \otimes b_{0+} \otimes b_{11} - \phi(b_{11}) \otimes b_{-0} \otimes b_{0+} - \phi(b_{0+}) \otimes b_{11} \otimes b_{-0} \\ & + \phi(b_{-1}) \otimes b_{1+} \otimes b_{00} + \phi(b_{00}) \otimes b_{-1} \otimes b_{1+} + \phi(b_{1+}) \otimes b_{00} \otimes b_{-1} \\ & + \phi(b_{00} + b_{11}) \otimes (b_{00} + b_{11}) \otimes (b_{00} + b_{11}). \\ & \phi(b_{xy}) = \sum_{j \in \{0,1\}^2} (b_x \otimes b_{j_1}) \otimes (b_{j_1} \otimes b_{j_2}) \otimes (b_{j_2} \otimes b_y) = T(\bullet) \\ & \phi(b_{00} + b_{11}) = \sum_{i \in \{0,1\}^3} (b_{i_1} \otimes b_{i_2}) \otimes (b_{i_2} \otimes b_{i_3}) \otimes (b_{i_3} \otimes b_{i_1}) = T(\bullet) \\ & f(T(\bullet))) \leq 6 \times 4 + 1 \times 7 = 31 < 2^5 \\ \hline \end{split}$$



31<32 ...

- works for all k-cycles $R(T(C_k)) \le 2^k 1$
- works for border rank as well
- and asymptotic rank without knowledge of the decomposition, always using the "7" mamu upper bound

$$\omega_k \le \omega_{k-2} + \omega_3$$

If $\omega = 2$, then $\omega_k = k - 1$ for all odd k

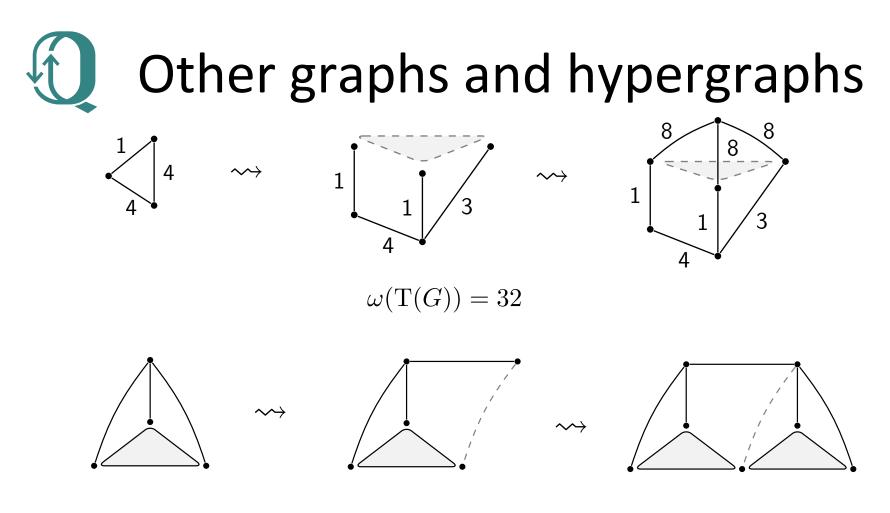
bigger than 0.3

(Le Gall)

=1 iff w=2

• uniformly bounded away from k

$$\omega_k \le k - \alpha \left(1 + \frac{1 - \alpha}{k - 1 + \alpha}\right) \le k - \alpha$$



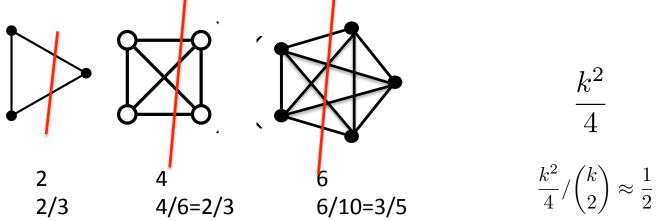
 $6 \le \omega(\mathbf{T}(H)) \le 6\omega/2.$

Works well for sparse graphs and hypergraphs! Your tensor? What about dense graphs?



Dense graphs

- Tensor surgery needs a good starting tensor The best we have is mamu!
- in tensor sugergy asymptotic log rank per edge increases (no problem for sparse graphs)

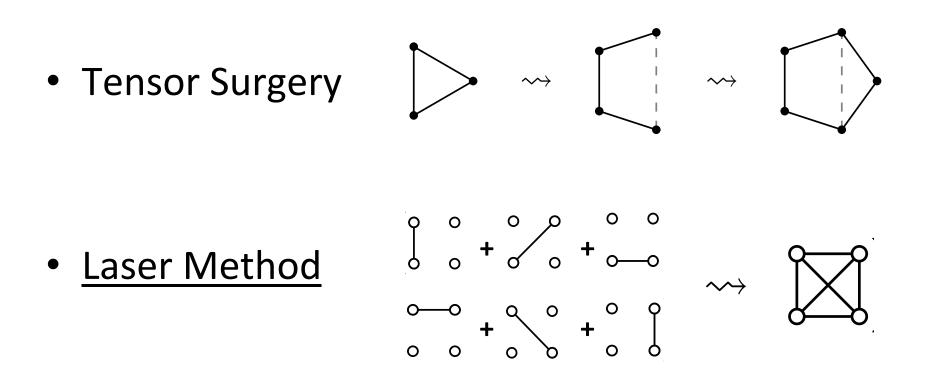


- best upper bounds from mamu covering $2.38/3 \approx 0.79$
- Can we beat this for some graph?



Overview

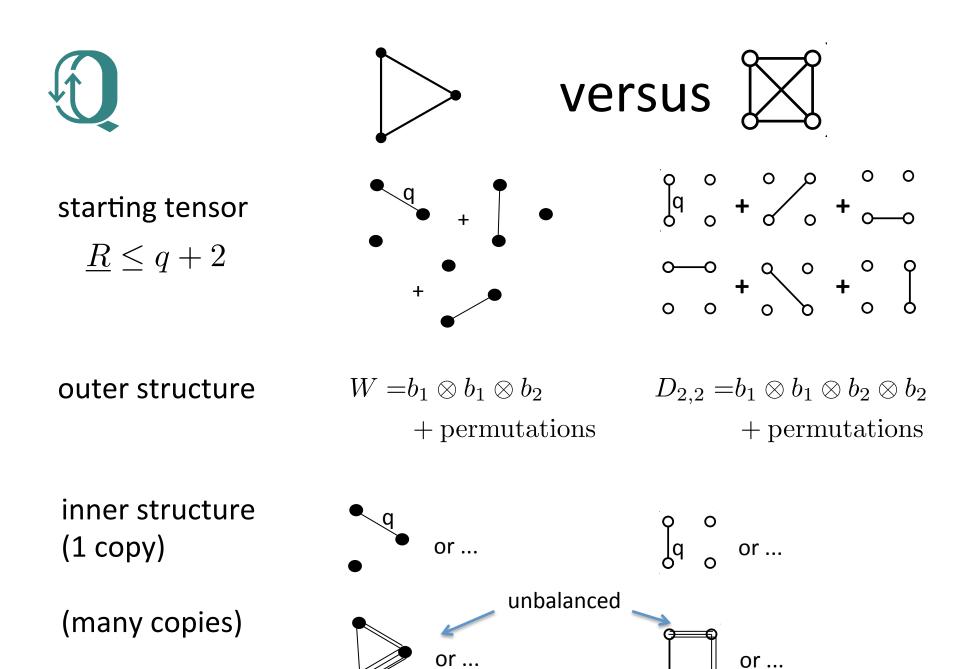
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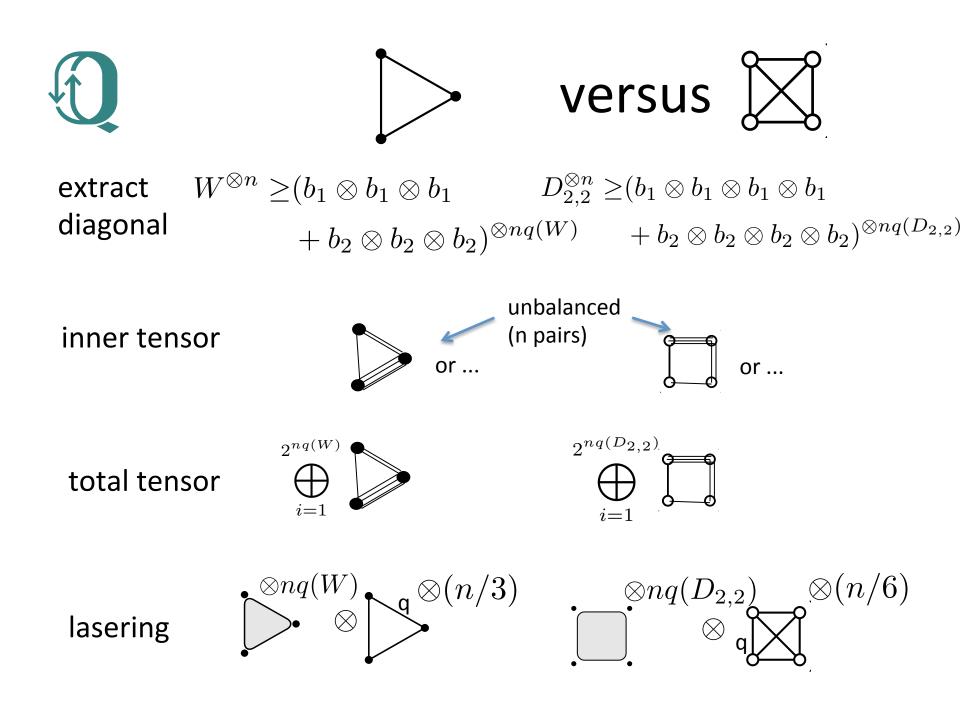


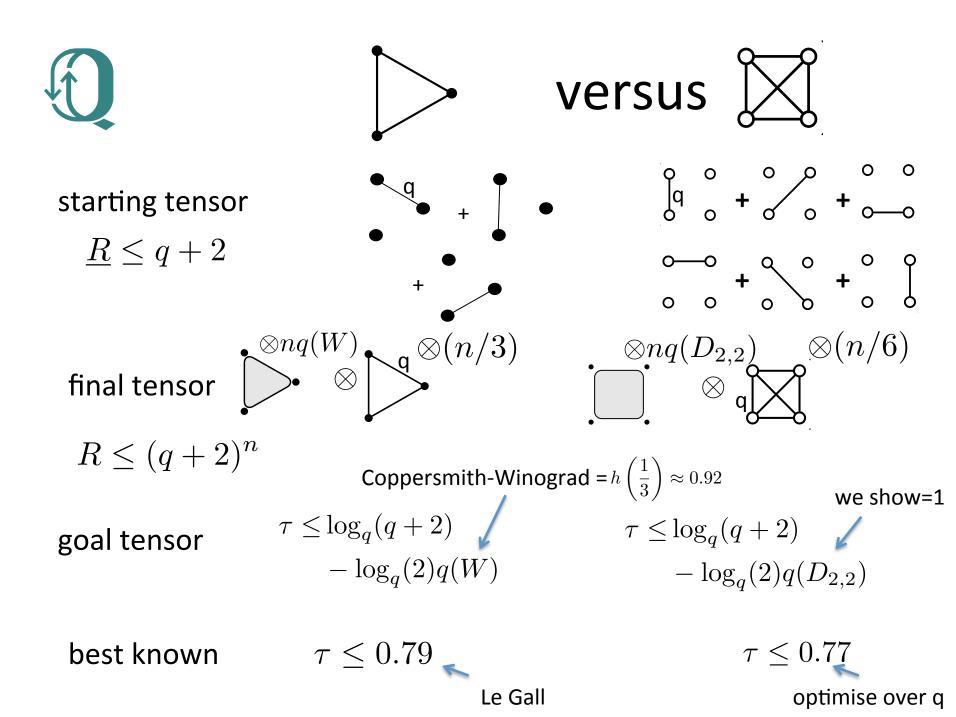


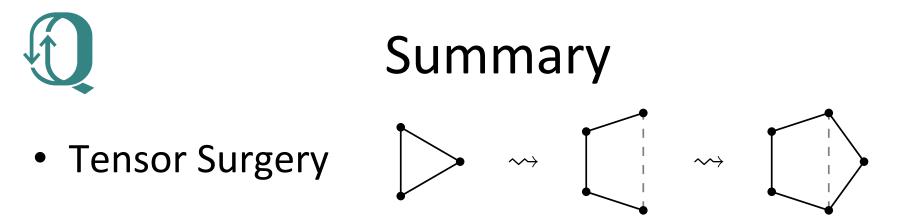
Laser method

- by Strassen for upper bound on mamu exponent
- with Coppersmith and Winograd starting tensor
- 1) choose starting tensor with low border rank with suitable coarse outer structure (W type) and fine inner structure (Mamu-type)
- 2) take many copies and extract unit tensors from W, each inner tensor is Mamu-type
- The Mamu-type tensors can be a little different. Schönhage's asymptotic sum inequality makes them equal (coherent – thus the name "laser")

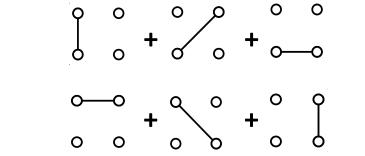








nontrivial rank results for k-cycle optimal asymptotic rank results good for sparse graphs



Laser Method beating matrix multiplication for best asymptotic rank per edge



Advertisement

- Tensor rank is not multiplicative under the tensor product
- with Asger Kjærulf-Jensen & Jeroen Zuiddam
- arxiv:1705.09379
- Main Results: $R(\underbrace{W \otimes W}_{6-tensor}) \leq 8 < 9 = R(W)^{2}$ $R(t^{\otimes n}) \leq poly(n)\underline{R}(t)^{n}$