

Quantum LDPC codes

Lecture 4

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Overview

- Brief overview of quantum Tanner code
- Conclusion: Should we abandon surface codes?

Quantum Tanner codes

Some constructions of QLDPC codes

Topological codes:

- 1997: Kitaev.
- 2002: Freedman, Meyer, Luo
- 2009: Bravyi, Poulin, Terhal bound:
constant $\times n$

$$d = \Omega(\sqrt{n}) \text{ but } k = 1.$$

$$d = \Omega\left(\sqrt{n\sqrt{\log n}}\right)$$

$$kd^2 \leq$$

Hypergraph-product codes and generalizations:

- 2009: Tillich, Zémor. HGP
with $k \propto n$
- 2013: Bravyi, Hastings. Homological products
- 2020: Hastings, Haah, O'Donnell: Fiber bundle codes_
- 2020: Panteleev, Kalachev: Lifted products
- 2020: Breuckmann and Eberhardt: Balanced products

$$d = \Omega(\sqrt{n})$$

$$d = \Omega\left(\frac{n^{\frac{3}{5}}}{\text{polylog}}\right)$$

$$d = n^{1-\varepsilon}/\log n$$

$$d = \Omega(n^{\frac{3}{5}})$$

Decoder for good LDPC codes

Good LDPC codes:

- 2021: Panteleev and Kalachev. Good QLDPC codes
- 2021: Dinur, Evra, Livne, Lubotzky, and Mozes. LTC codes
- 2022: Leverrier, Zémor

Linear time decoders for quantum Tanner codes:

- 2022: Gu, Pattison, Tang
- 2022: Dinur, Hsieh, Lin, Vidick
- 2022: Leverrier, Zémor

Left-Right Cayley Complex

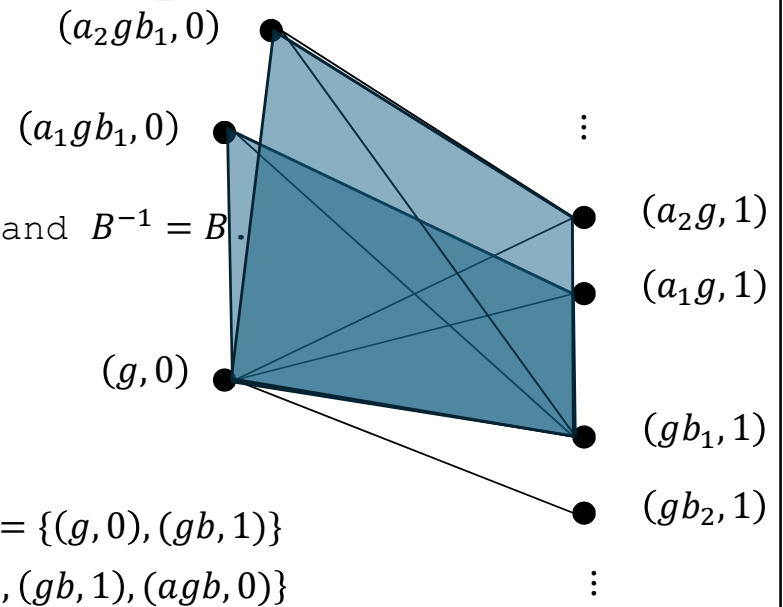
Input:

- A finite group G
- Two subsets A, B of G such that $A^{-1} = A$ and $B^{-1} = B$.

Output:

A complex $X = (V, E, F)$ where:

- $V = V_0 \cup V_1$ where $V_i = G \times \{i\}$
- $E = E_A \cup E_B$ where $E_A = \{(g, 0), (ag, 1)\}$ and $E_B = \{(g, 0), (gb, 1)\}$
- $F =$ set of squares $f(g, a, b) := \{(g, 0), (ag, 1), (gb, 1), (agb, 0)\}$

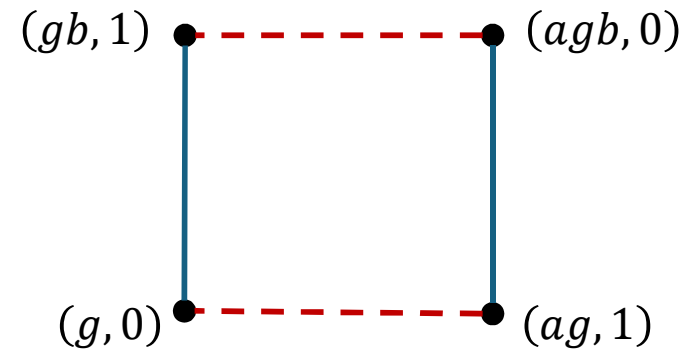
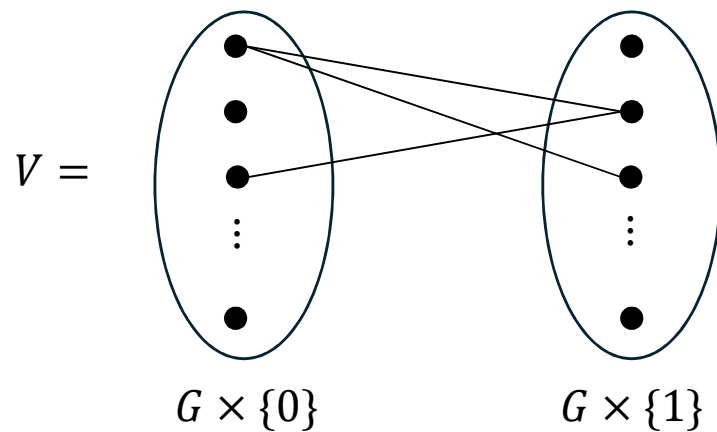


Local structure

$$G = \{g_1, g_2, \dots\}$$

$$A = \{a_1, a_2, \dots\} \subseteq G$$

$$B = \{b_1, b_2, \dots\} \subseteq G$$



Type A edge:

$$a \times \cdot$$

Type B edge:

$$\cdot \times b$$

—————

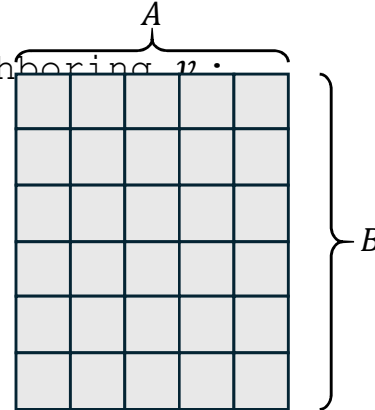
Neighborhood of a vertex

Consider $v = (g, 0)$

- Neighboring vertices: $(ag, 1)$ with $a \in A$ and $(gb, 1)$ for $b \in B$
- Neighboring faces: $\{(g, 0), (ag, 1), (gb, 1), (agb, 0)\}$ for each pair $(a, b) \in A \times B$.

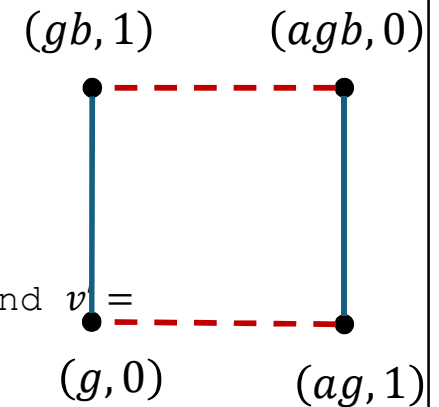
Abstract representation the faces neighboring v .

$$F(v) = A \times B =$$



Intersection of two neighborhoods

Question. What is $F(v) \cap F(v')$ for two vertices $v = (g, 0)$ and $v' = (g', 1)$?



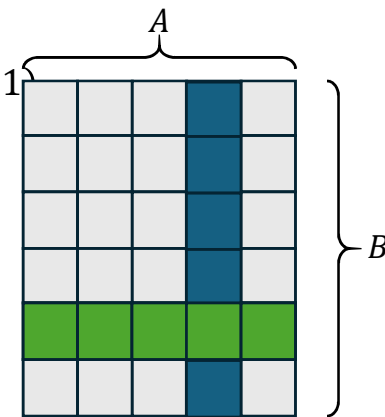
- If $F(v) \cap F(v') \neq \emptyset$, then v and v' must share an edge.
- Why?

- If they share a type A edge: $v = (g, 0)$ and $v' = (ag, 1)$
- Then $F(v) \cap F(v')$ is the set of faces:

$$\{(g, 0), (ag, 1), (gb, 1), (agb, 0)\} \text{ for } b \in B$$

$$F(v) \cap F(v') = B =$$

- If they share a type B edge:



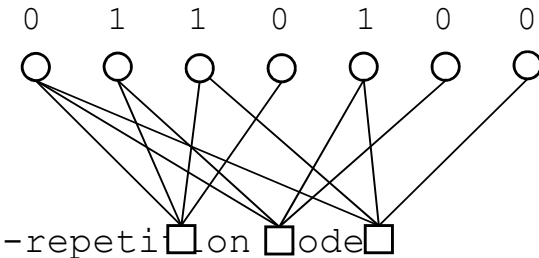
Tensor codes of two classical codes

Def. A **codeword** of $C_1 \otimes C_2$ is bitstring forming a $n_1 \times n_2$ matrix x such that

- each column of x is in C_1 ,
- each row of x is in C_2 ,

Ex.

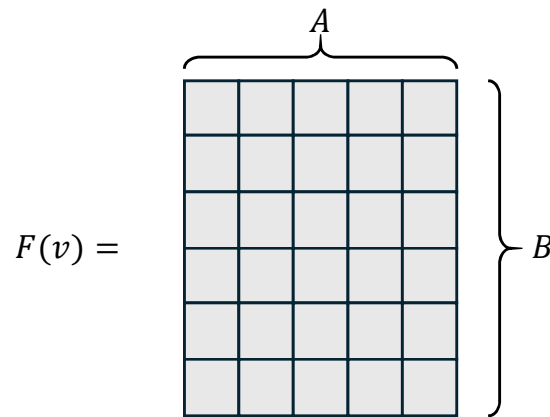
0	1	1	0	1	0	0
0	1	1	0	1	0	0
0	1	1	0	1	0	0



is in the product of the Hamming code and the 3-repetition code

Prop. The code $C_1 \otimes C_2$ has parameters $[n_1 n_2, k_1 k_2]$.

Definition of quantum Tanner codes



- Select a finite group G .
- Select $A, B \subseteq G$ such that $A^{-1} = A$ and $B^{-1} = B$.
- Select two codes C_A and C_B with length $|A|$ and $|B|$.
- Place a **qubit** on each face of the left-right Cayley complex (V, E, F)
- For each $v = (g, 0)$, for each $c \in C_A \otimes C_B$, define a **X generator** on $F(v)$ acting on the support of c .
- For each $v = (g, 1)$, for each $c \in C_A^\perp \otimes C_B^\perp$, define a **Z generator** on $F(v)$ acting on the support of c .

Prop. The generators commute because $F(v) \cap F(v')$ is either empty, or a row of $F(v)$ or a column of $F(v)$.

The sets must satisfy the TNC condition for all a, g, b : $ag \neq gb$

Example of Cayley graphs

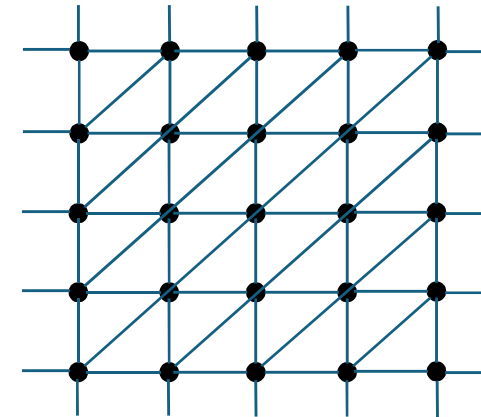
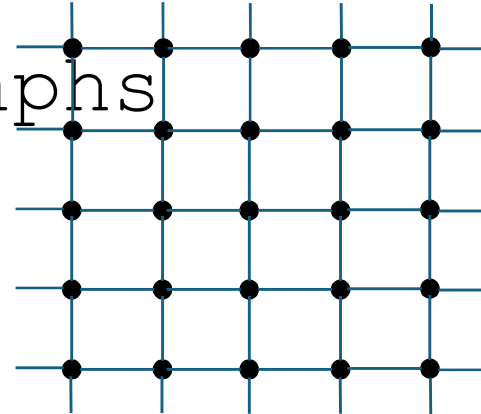
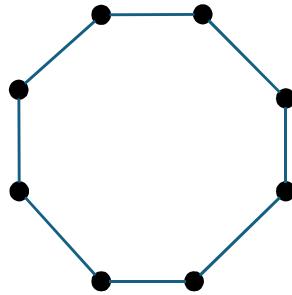
- $G = \mathbb{Z}$,
- $A = \{\pm 1\}$



Example of Cayley graphs

- $G = \mathbb{Z}^2, A = \{(\pm 1, 0), (0, \pm 1)\}$
- $G = \mathbb{Z}^2, A = \{(\pm 1, 0), (0, \pm 1), (\pm 1, \pm 1)\}$
- $G = \mathbb{Z}_8, A = \{\pm 3\}$

Why do we need $A = A^{-1}$?



Double cover of a Cayley graph

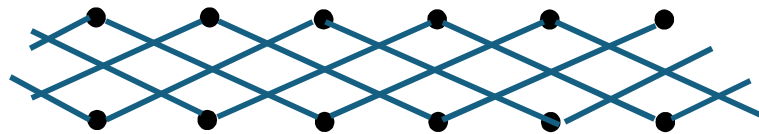
It is the graph with

- $V = V_0 \cup V_1$ with $V_0 = G \times \{0\}$ and $V_1 = G \times \{1\}$
- Two types of edges: $\{(g, 0), (ag, 1)\}$ and $\{(g, 1), (ag, 0)\}$

Ex. $G = \mathbb{Z}, A = \{\pm 2\}$

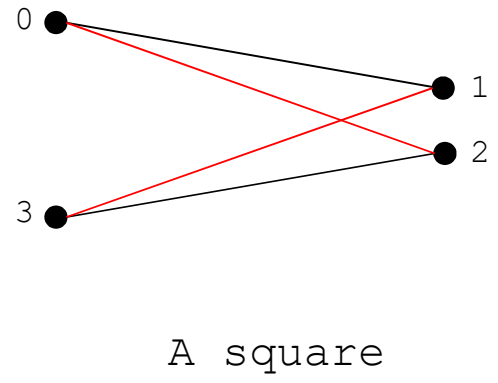
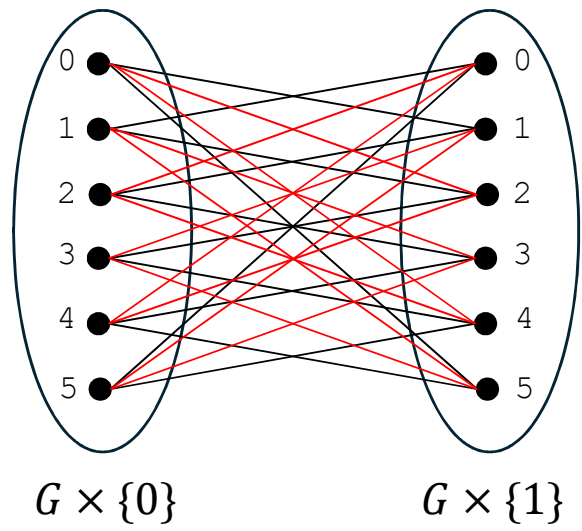
$G \times \{0\} =$

$G \times \{1\} =$



Example

$$G = \mathbb{Z}_6, A = \{\pm 1\}, B = \{\pm 2\}$$



Example - $G = \mathbb{Z}_9$

$G = \mathbb{Z}_9, A = \{\pm 1\}, B = \{\pm 2\}$

Type A edge:

(+1, flip)



(-1, flip)



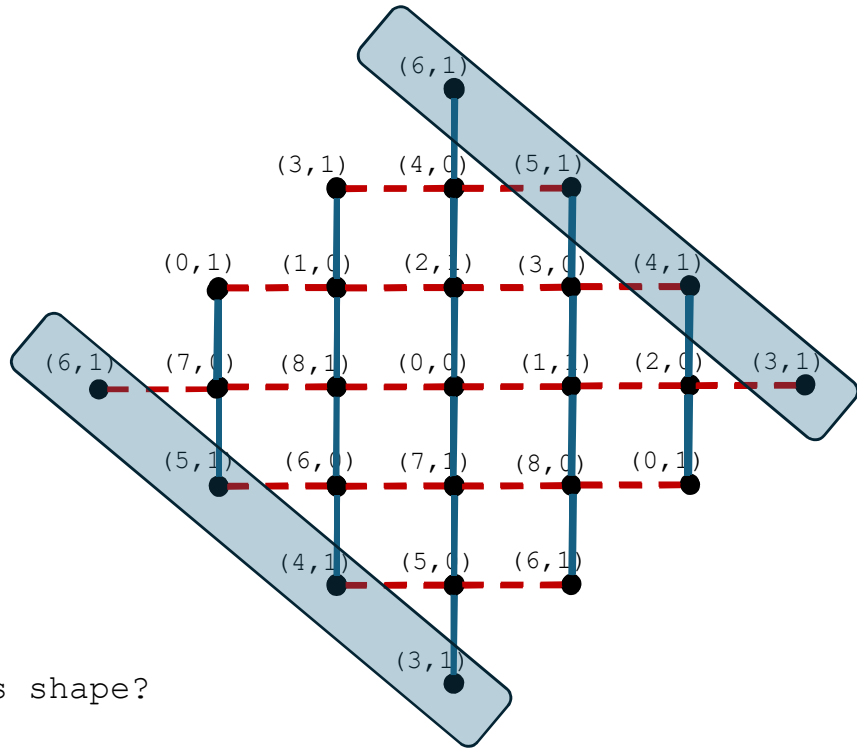
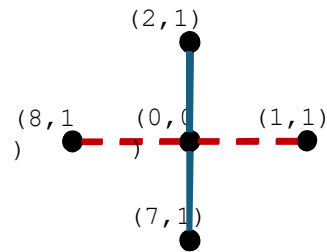
Type B edge:



(+2, flip)



(-2, flip)



What is this shape?

Example - $G = \mathbb{Z}_3 \times \mathbb{Z}_3$ $G = \mathbb{Z}_3 \times \mathbb{Z}_3, A = \{(\pm 1, 0)\}, B = \{(0, \pm 1)\}$

Type A edge:

$(+1, 0, \text{flip})$



$(-1, 0, \text{flip})$



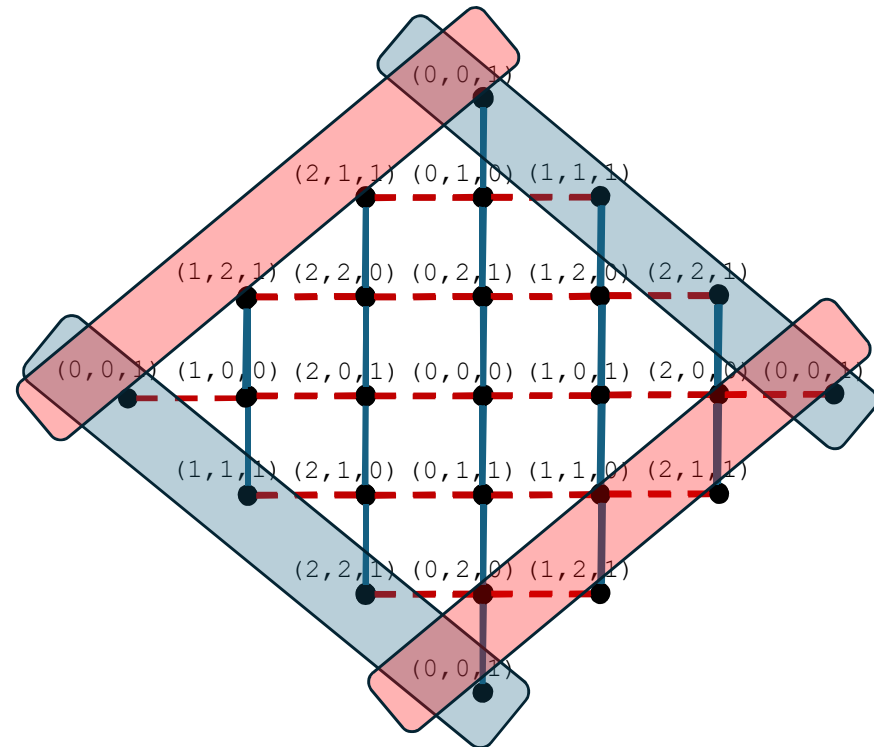
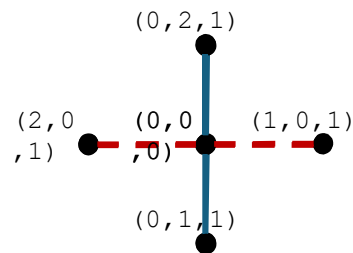
Type B edge:



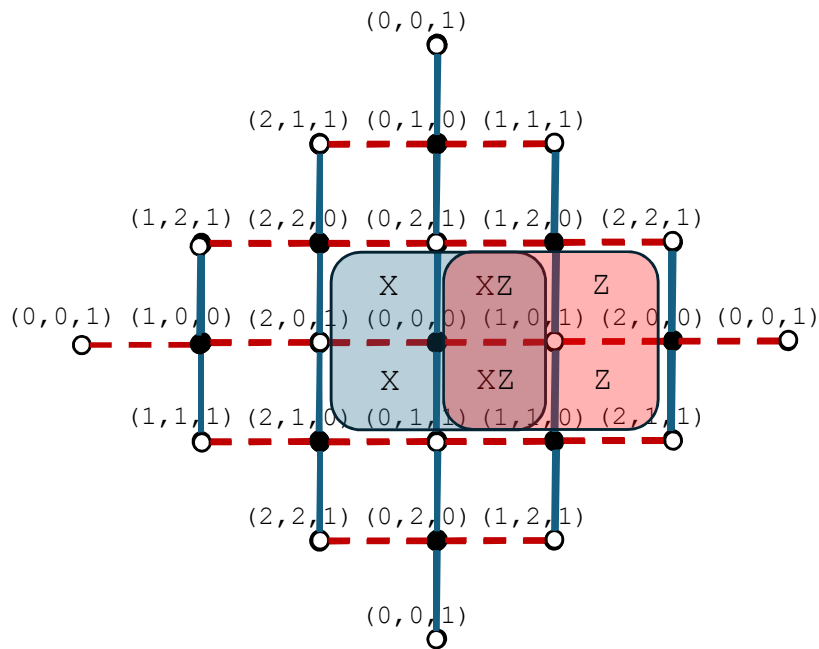
$(+1, 0, \text{flip})$



$(-1, 0, \text{flip})$



Example - $G = \mathbb{Z}_3 \times \mathbb{Z}_3$



We recover the toric code.

Construction:

- Qubits are on faces
 - For $v = (g, 0)$, $c \in C_A \otimes C_B$, define a **X generator** on $F(v)$ acting on the support of c .
 - For $v = (g, 1)$, $c \in C_A^\perp \otimes C_B^\perp$, define a **Z generator** on $F(v)$ acting on the support of c .
- $(g, 0) = \bullet$ X stabilizer generators

$F(v) = A \times B =$

We need to a tensor code $C_A \otimes C_B$ on $A \times B$:
 $\Rightarrow C_A = C_B = \{00, 11\}$

$C_A \otimes C_B =$

0	0	1	1
0	0	1	1

\Rightarrow X stabilizer generator:

X	X
X	X

How to get good LDPC codes

Take:

- $G = \text{PSL}_2(q^i)$
- $C_A = \text{random code}$
- $C_B = \text{random code}$

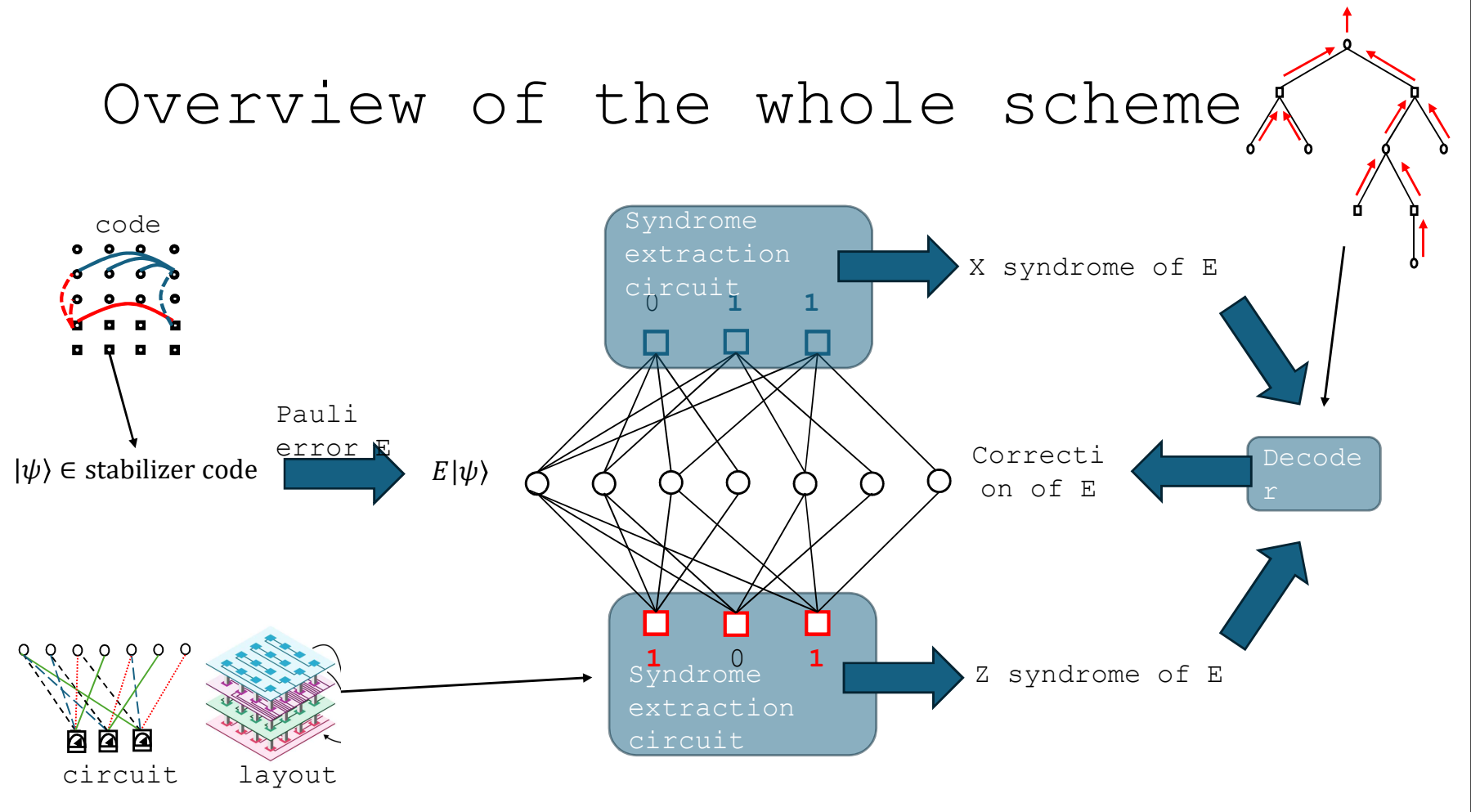
This leads to a family of good quantum LDPC codes.

Question:

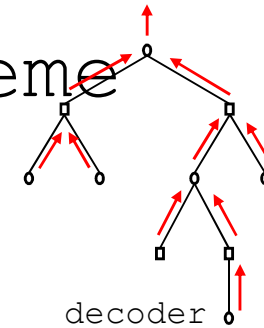
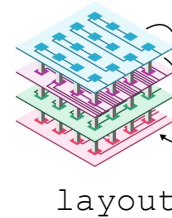
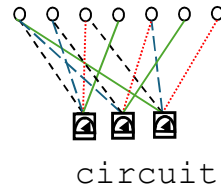
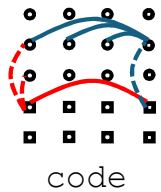
Should we all replace our codes by good quantum LDPC codes?

Conclusion

Overview of the whole scheme



Overview of the whole scheme



Select two random Tanner graphs with

- 4s bits with degree 3.
- 3s checks with degree 4.
- girth ≥ 8

Construct their HGP.

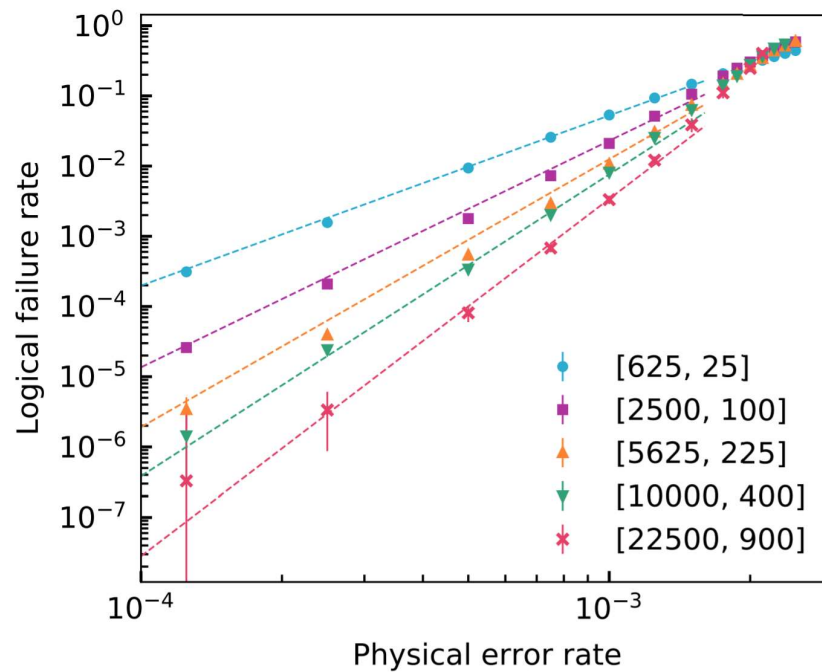
- Estimate the performance of the HGP code using the SSF decoder.
- Select the best HGP

- Compute the Tanner graph T of the code.
- Compute an edge coloration of T
- Construct the color-based syndrome extraction circuit.

- Compute a layered decomposition of the Tanner graph.
- Compute an edge coloration of T
- Construct the color-based syndrome extraction circuit.

- We use BP for decoding in a single shot manner.
- We estimate the logical error rate over 10 rounds of syndrome extraction.
- To check if a logical error occurs, we use SSF decoder to correct the residual

Numerical results



Noise threshold:

0.28% (instead of 0.7% for surface codes)

physical qubits per logical qubit:

49 (instead of thousands for surface codes)

Logical failure rate	10^{-9}	10^{-12}	10^{-15}
Logical qubits	1600	6400	18 496
Surface code physical qubits	387 200	2 880 000	13 354 112
HGP code physical qubits	78 400	313 600	906 304
Improvement using HGP codes	4.94×	9.18×	14.73×

Is our scheme fault-tolerant?

Syndrome extraction:

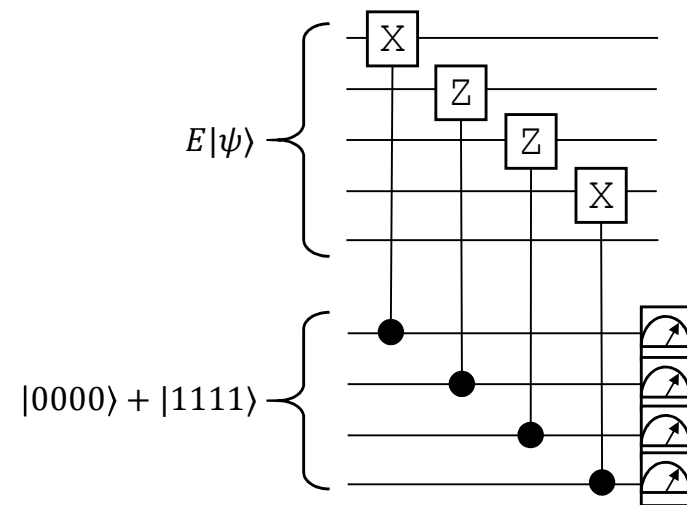
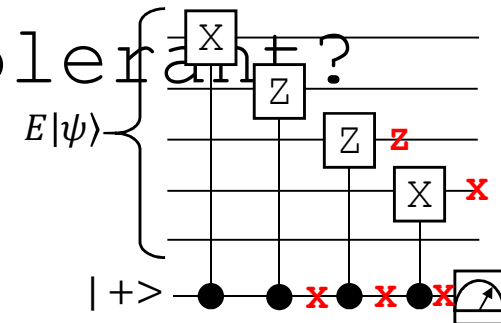
- What is this circuit?
- Are there 'bad' faults for this circuit?
- What is the effect of a X fault on the ancilla qubit?
- Can we avoid that?
- Should we avoid that?

Decoder:

- BP corrects each qubits independently based on marginal probability. Is it a problem?
- The decoder uses noisy syndrome data. Is it a problem?

Conclusion:

- Practical FT \neq Theoretical FT



What could be improved?

1. Improve the code:

- linear distance,
- reduced code length,
- denser family.

2. Improve the circuit:

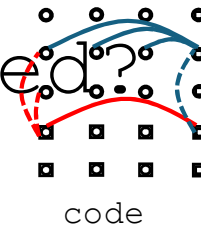
- the circuit is not FT (reduced distance).

3. Improve the decoder:

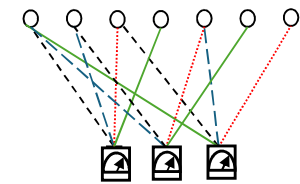
- better logical error rate,
- linear complexity,
- hardware optimization.

4. Improve the simulation:

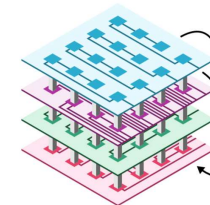
- Refine the numerical estimate of the logical error rate.
- Simulate longer lifetime.



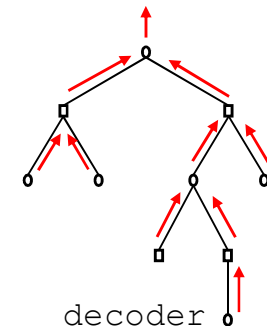
code



circuit



layout



decoder

What about computation?

- 2013: Gottesman - Fault-Tolerant Quantum Computation with constant overhead.
- Other proposals for fault-tolerant logical gates in QLDPC codes^{1,2,3,4}.

(c)

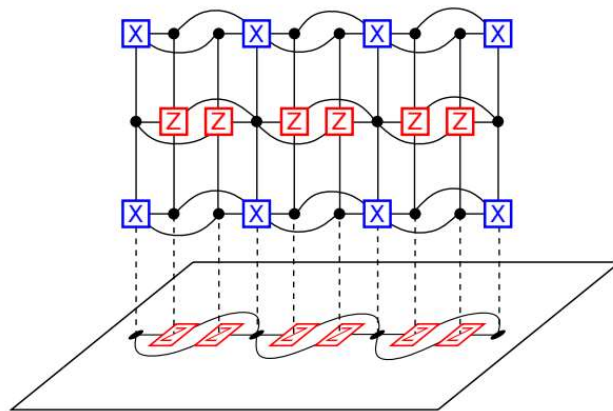
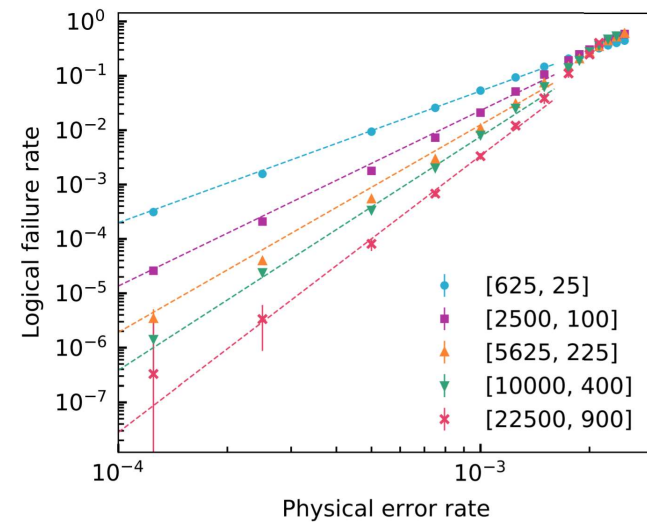


Figure from ref 3.

1. Jochym-O'Connor, arxiv:1807.09783
2. Krishna, Poulin, arxiv:1909.07424
3. Cohen, Kim, Bartlett, Brown, arxiv: 2110.10794
4. Breuckmann, Burton, arxiv:2202.06647

What could go wrong with this approach?

- Noise could be correlated.
- Noise could be non-Pauli.
- We may be unable to build sufficiently reliable hardware (need $p = 10^4$).
- The blocks could be too big (we need $n = 900,000$).
- The decoder could be too slow.
- Fault-tolerant operations may be expensive.
- We may be unable to build the required long-range connections.
- We may need too many long-range connections.
- Building insulated layers of long-range connections may be hard.



Logical failure rate	10^{-9}	10^{-12}	10^{-15}
Logical qubits	1600	6400	18 496
Surface code physical qubits	387 200	2 880 000	13 354 112
HGP code physical qubits	78 400	313 600	906 304
Improvement using HGP codes	4.94×	9.18×	14.73×

Other encoding strategies

Now what?

Should we abandon surface codes?

Hyperbolic codes:

- Higgott, Breuckmann (2020)
- Breuckmann, Vuillot, Campbell, Krishna, Terhal (2017)

Floquet codes:

- Haah, Hastings (2021)

Spacetime code:

- Delfosse, Paetznick (2023)

Appendix - History of
classical LDPC codes

Brief (and biased) history

- Asymptotic results: Shannon (1940's)
 - Channel capacity: If we use a channel an infinite number of times, what is the maximum number of bits of information that we can send per use of the channel?
 - Basic idea: Random codes are optimal.
 - Problem: How to encode? How to decode?
- Birth of modern coding: Elias and Gallager (1960's)
 - Linear codes perform as well as random codes but encoding is easy.
 - Convolutional codes too.
 - Erasure channel as a toy model.
 - LDPC codes.

Brief (and biased) history

- The comeback of modern coding: Berrou, Mackay, Neal, Richardson, Urbanke, Shorkollahi, ... (1990's):
 - Turbo codes
 - LDPC codes
 - Decoding is easy! But is it optimal?
- Capacity achieving codes.
 - Proofs for the erasure channel first.
 - Irregular LDPC codes.
 - Spatially-coupled codes achieve capacity (Kudekar, Richardson, Urbanke 2013).
- Today LDPC codes are in your cell phone, your laptop, your WiFi...

How far are we from the classical story?

Classical goal:

- Achieve capacity with a linear time decoder.

Def. The capacity of a channel N is the maximum rate $\frac{k}{n}$ of a family of codes with vanishing logical error rate.

Quantum goal:

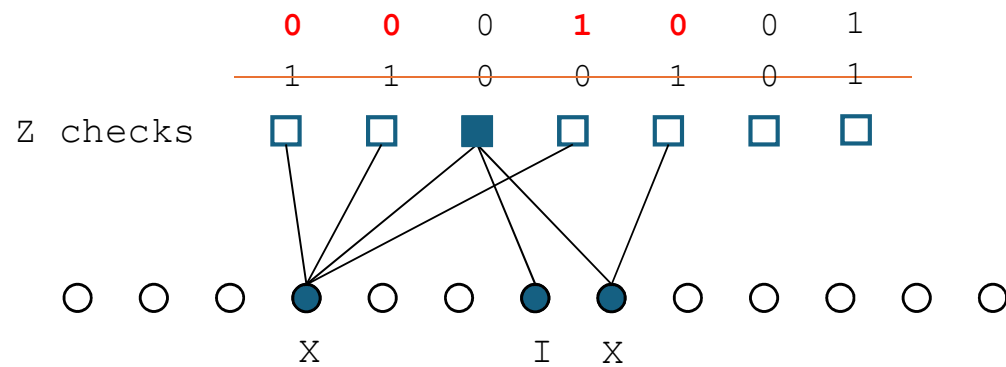
- Achieve capacity?
- Build a fault-tolerant quantum computer?

We can still learn from the classical case: For instance, starting with the erasure channel.

Appendix - Small Set Flip (SSF) decoder

Leverrier, Tillich, Zémor -
[arXiv:1504.00822](https://arxiv.org/abs/1504.00822) (2015)

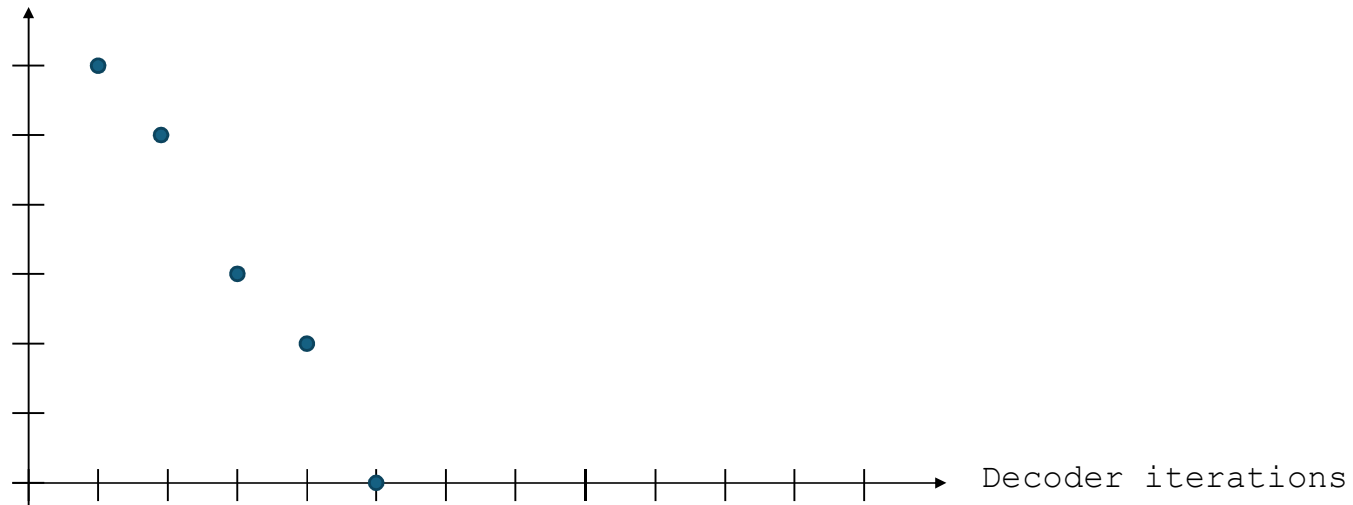
Small Set Flip decoder - Basic idea



1. Select a Z check
2. Select an error E_X inside the Z check that reduces the syndrome weight.
3. Update the syndrome

Small Set Flip decoder - Basic idea

Syndrome weight



Leverrier, Tillich, Zémor -
arXiv:1504.00822 (2015)

Small set flip decoder

Def: A *critical generator* g is a Z stabilizer generator that contains a X error that reduces the weight of the syndrome.

Input: A syndrome value $s_c = 0$ or 1 for each Z check node.

Output: A correction for X errors.

1. While there exists a critical generator g :
2. Select an error $E_X(g)$ included in g such that $\frac{\text{weight reduction}}{|E_X(g)|}$ is maximum.
3. Update the syndrome: Add $s(E_X(g))$ to the syndrome.
4. Return the product of all the $E_X(g)$.

Theorem. [Leverrier, Tillich, Zémor, 2015]. Under some conditions about the expansion of the two input graphs, *the small set flip decoder for HGP codes corrects any set of*

Small set flip decoder - Complexity

Complexity: $O(2^w n)$

Notation:

- n = code length.
- w = max. degree of a check (max. weight of a stabilizer generator).

Remark:

- Linear-time complexity for bounded degree Tanner graphs.
- May be slow in practice for large w .