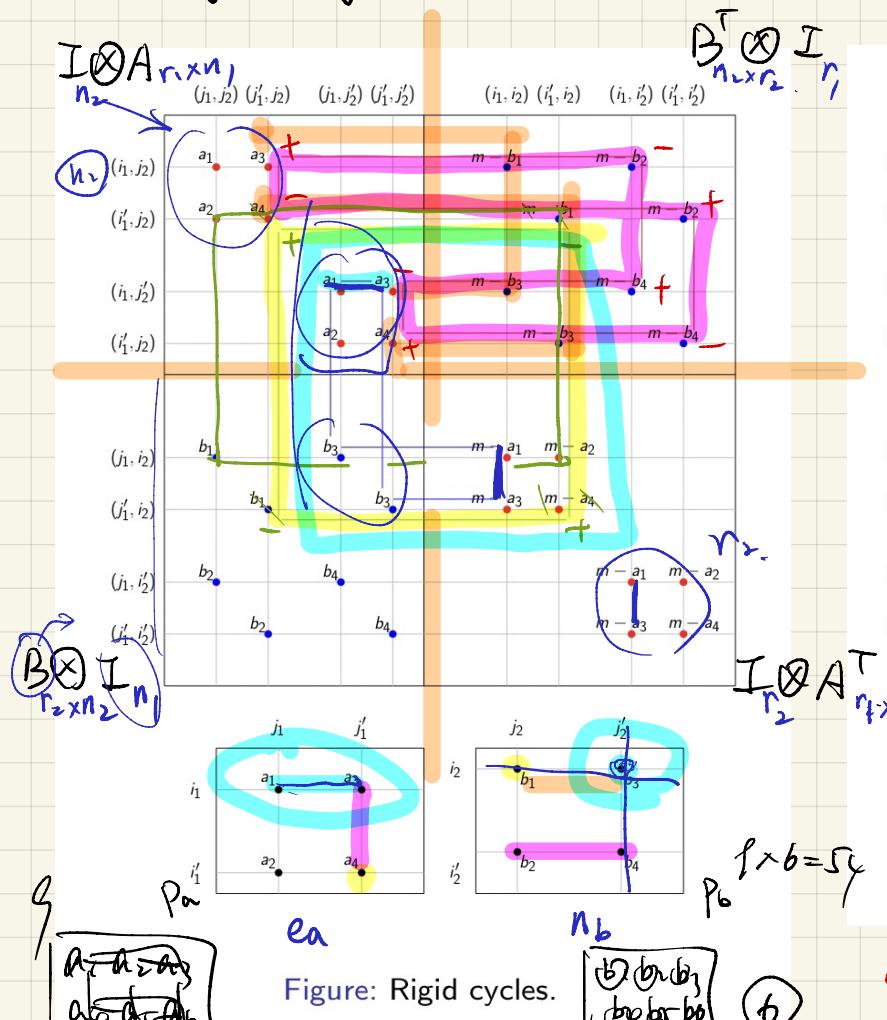


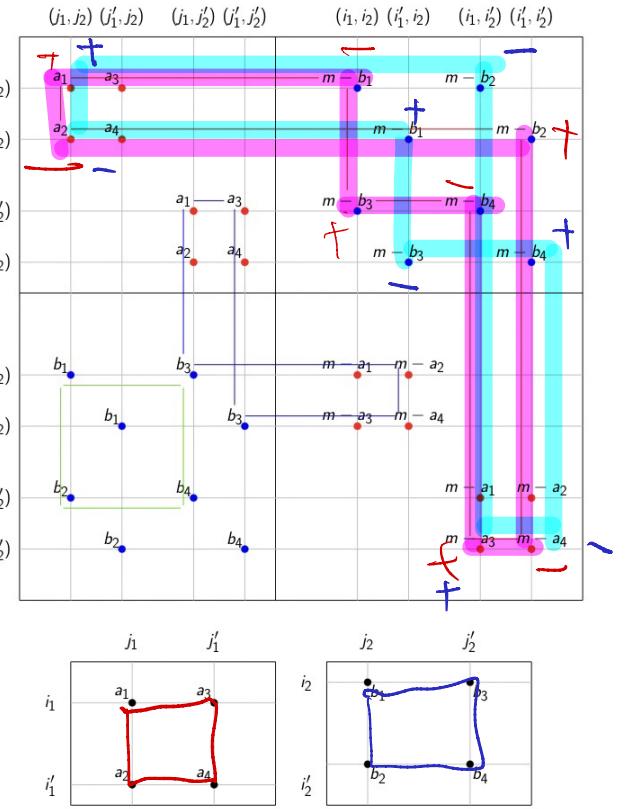
# SC-QLDPC Cycles.

## ① Cycle Optimization

### Rigid Cycles.



### Flexible Cycles.



$$a_1 - (m - b_1) + (m - b_3) - (m - b_4) + (m - a_3)$$

$$- (m - a_4) + (m - b_2) - a_2$$

$$= (a_1 - a_2 + a_4 - a_3) + (b_1 - b_2 + b_4 - b_3)$$

$$a_1 - (m - b_2) + (m - a_3) - (m - a_4) + (m - b_2)$$

$$- (m - b_3) + (m - b_1) - a_2$$

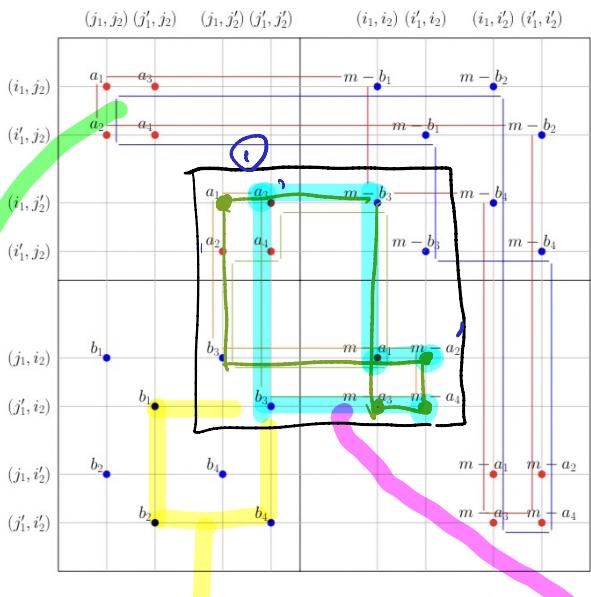
$$= (a_1 - a_2 + a_4 - a_3) - (b_1 - b_2 + b_4 - b_3)$$

$$\text{If } S = |a_1 - a_2 + a_4 - a_3| = |b_1 - b_2 + b_4 - b_3|$$

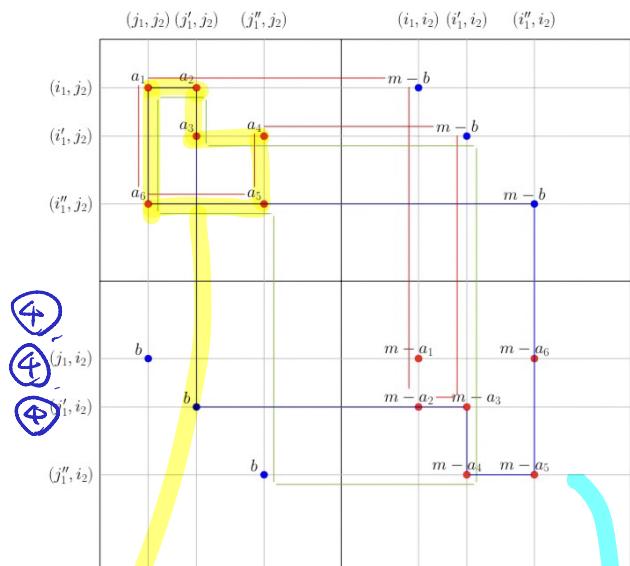
$$S = 0 \Rightarrow 2 \text{ cycles} - 8$$

$$S \neq 0 \Rightarrow 1 \text{ cycle} - 8$$

Independent of  $a_i, b_i$ .



(a) Cycles-4 induced flexible cycle candidates.



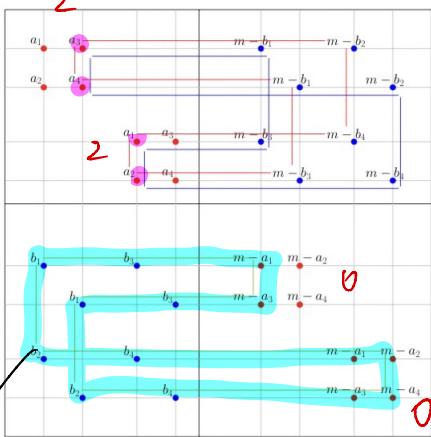
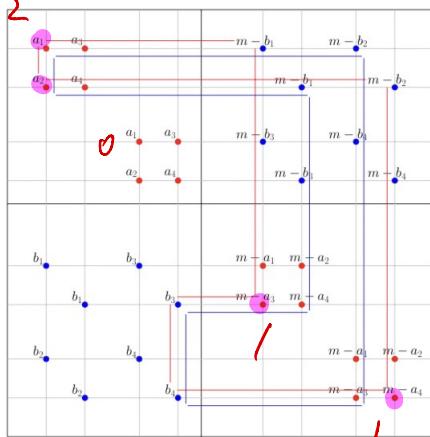
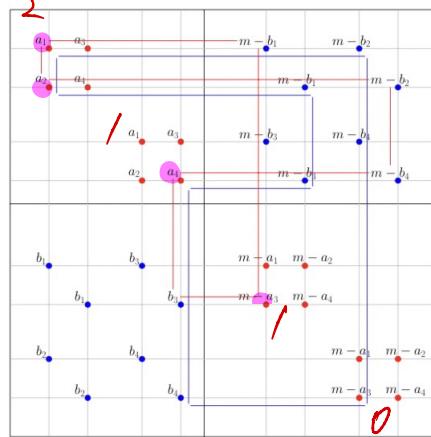
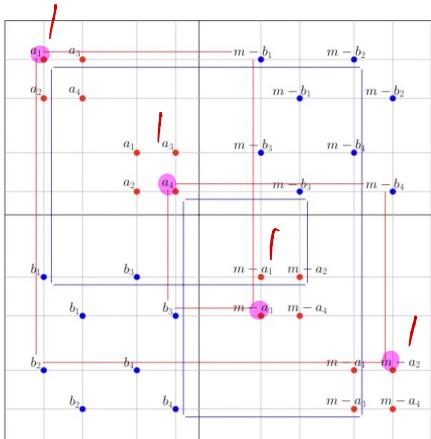
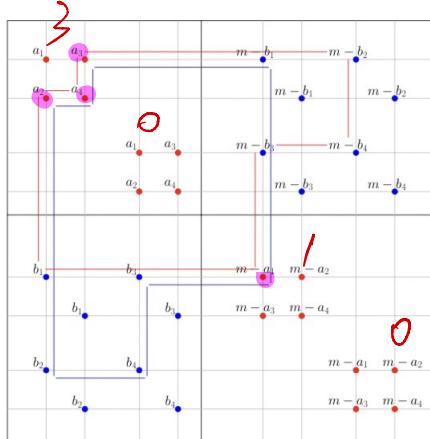
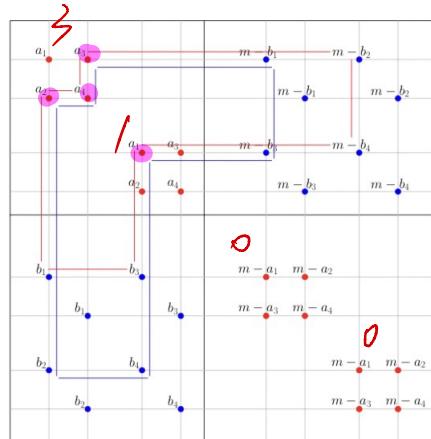
(b) Cycles-6 induced flexible cycle candidates.

$$\begin{aligned}
 N_4 &= n_{4,0,0}^A(n_2 + r_2) + n_{4,0,0}^B(n_1 + r_1), \\
 N_6 &= n_{6,0,0}^A(n_2 + r_2) + n_{6,0,0}^B(n_1 + r_1) + 12(n_{4,0,0}^A e_2 + n_{4,0,0}^B e_1), \\
 N_8 &= n_{8,0,0}^A(n_2 + r_2) + n_{8,0,0}^B(n_1 + r_1) + 30(n_{6,0,0}^A e_2 + n_{6,0,0}^B e_1) \\
 &\quad + 124 \left( 2n_{4,0,0}^A n_{4,0,0}^B + \sum_{(i,j) \neq (0,0)} (n_{4,i,j}^A + n_{4,-i,-j}^A)(n_{4,i,j}^B + n_{4,-i,-j}^B) \right) \cdot 4
 \end{aligned}$$

$n_{sg, i, j}^A (n_{sg, i, j}^B) =$  number of closed paths of length  $2g$  with alternating sum  $i, j$  in 1st/2nd dimension.

$$f^A(S, T) = \sum_{i=0}^{m_1} \sum_{j=0}^{m_2} P_{i,j}^A S^i T^j, \quad f^B(S, T) = \sum_{i=0}^{m_1} \sum_{j=0}^{m_2} P_{i,j}^B S^i T^j$$

$$\begin{aligned}
 \mathbb{E}_{p^A, p^B}[N_6] &= n_6^A(n_2 + r_2) \left[ (f^A(S, T) f^A(S^{-1}, T^{-1}))^3 \right]_{0,0} + n_6^B(n_1 + r_1) \left[ (f^B(S, T) f^B(S^{-1}, T^{-1}))^3 \right]_{0,0}, \\
 \mathbb{E}_{p^A, p^B}[N_8] &= n_8^A(n_2 + r_2) \left[ (f^A(S, T) f^A(S^{-1}, T^{-1}))^4 \right]_{0,0} + n_8^B(n_1 + r_1) \left[ (f^B(S, T) f^B(S^{-1}, T^{-1}))^4 \right]_{0,0} \\
 &\quad + 30 \left( n_6^A e_B \left[ (f^A(S, T) f^A(S^{-1}, T^{-1}))^3 \right]_{0,0} + n_6^B e_A \left[ (f^B(S, T) f^B(S^{-1}, T^{-1}))^3 \right]_{0,0} \right) \\
 &\quad + 248 n_4^A n_4^B \left[ (f^A(S, T) f^A(S^{-1}, T^{-1}))^2 (f^B(S, T) f^B(S^{-1}, T^{-1}))^2 \right]_{0,0}.
 \end{aligned}$$

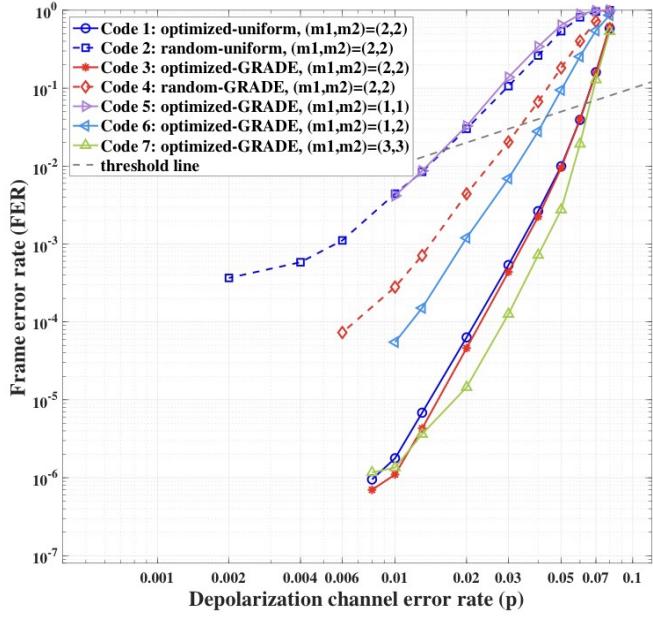
(a)  $\{2, 2\}, \{0, 0\}\}$ .(b)  $\{2, 0\}, \{1, 1\}\}$ .(c)  $\{2, 1\}, \{1, 0\}\}$ .(d)  $\{1, 1\}, \{1, 1\}\}$ .(e)  $\{3, 0\}, \{1, 0\}\}$ .(f)  $\{3, 1\}, \{0, 0\}\}$ .

$$\delta_1 - (m - a_1) + (m - a_3) - \delta_X + \delta_2 - (m - a_4) + (m - a_2) - \delta_Z$$

$= a_1 - a_2 + a_4 - a_3 \rightarrow$  induced by a single cycle-4 in Pan or Pb. We didn't count this because cycles-4 will be removed anyways.

## Optimization

- ① Start with a random pair of permutation matrix
- ② Modify each value to the value that locally minimizes the number of cycles (or weighted sum of  $N_b$  &  $N_g$ )  
 → first priority: remove all cycles-4.



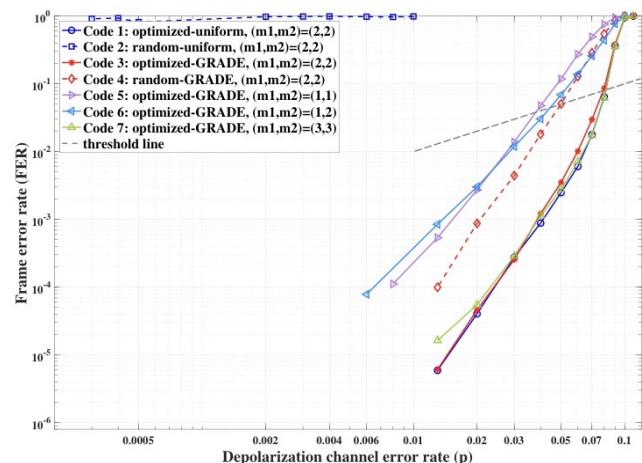
**Table:** Number of Flexible Cycles in [[7300, 2500]] codes

Common parameters		$(r_1, n_1, r_2, n_2, L_1, L_2) = (3, 8, 3, 8, 10, 10)$					
Codes	Code 1	Code 2	Code 3	Code 4	Code 5	Code 6	Code 7
Distributions	Uniform			GRADE			
Optimized by GRADE	✓	✗	✓	✗	✓	✓	✓
$(m_1, m_2)$	(2, 2)			(1, 1)			(1, 2)   (3, 3)
Number of cycles 4	0	110	0	66	0	0	0
Number of cycles 6	11	264	11	143	583	198	0
Number of cycles 8	5113	48142	5131	23120	64585	23266	1144

has cycles 4

## Parameters:

- $r = r_1 n_2 + r_2 n_1 = 48$
- $n = n_1 n_2 + r_2 r_1 = 73$
- $k = n - r = 25$
- $N = n L_1 L_2 = 7300$
- $K = k L_1 L_2 = 2500$



**Table:** Number of Flexible Cycles in [[5800, 1600]] codes

Common parameters		$(r_1, n_1, r_2, n_2, L_1, L_2) = (3, 7, 3, 7, 10, 10)$					
Codes	Code 1	Code 2	Code 3	Code 4	Code 5	Code 6	Code 7
Distributions	Uniform			GRADE			
Optimized by GRADE	✓	✗	✓	✗	✓	✓	✓
$(m_1, m_2)$	(2, 2)			(1, 1)			(1, 2)   (3, 3)
Number of cycles 4	0	70	0	40	0	0	0
Number of cycles 6	0	70	0	140	320	70	0
Number of cycles 8	2316	>17708	2466	>18710	30424	8594	380

## Parameters:

- $r = r_1 n_2 + r_2 n_1 = 42$
- $n = n_1 n_2 + r_2 r_1 = 58$
- $k = n - r = 16$
- $N = n L_1 L_2 = 5800$
- $K = k L_1 L_2 = 1600$

# Decoder

## ① Binary (only for CSS)

- Split X, Z errors. ex.  $e_Z$ .

- Decode  $e_X, e_Z$  by  $H_X, H_Z$  using binary BP decoder, respectively.

## ② Non-binary (also work for non-CSS codes)

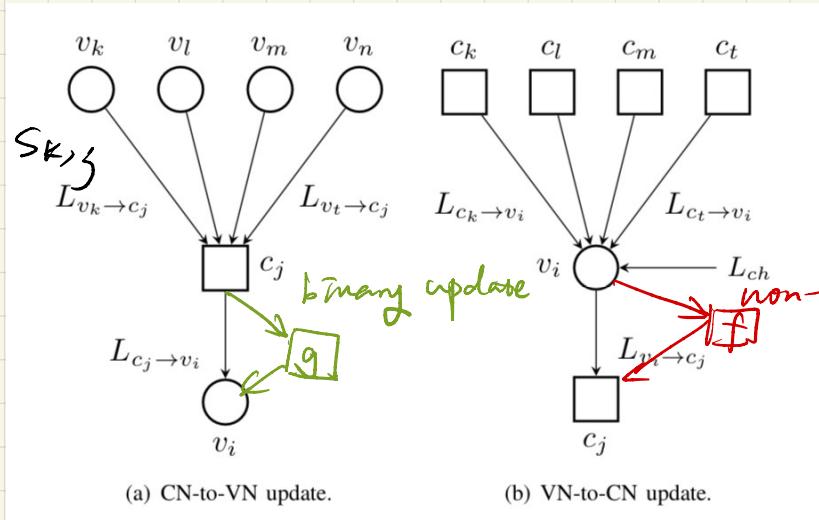
-  $\{I, X, Y, Z\} \rightarrow \{0, 1, 2, 3\}$

$e \in \{0, 1, 2, 3\}^N$ ,  $H \in \{0, 1, 2, 3\}^{(M+K) \times N}$

- Syndrome  $S \in \{0, 1\}^N$ .

$$\log \frac{\Pr[S_{k,j}, v_k >_S 0]}{\Pr[S_{k,j}, v_k >_S 1]} \rightarrow$$

$$\log \frac{P_x}{P_z}, \log \frac{P_Y}{P_Z}, \log \frac{P_Z}{P_Z}) \rightarrow$$



$$g(l; s) = \begin{cases} (0, -l, -l), & S=X \\ (-l, 0, -l), & S=Y \\ (-l, -l, 0), & S=Z \end{cases}$$

$$f((l_x, l_y, l_z); s) = \begin{cases} \log(1 + e^{l_x}) - \log(e^{l_y} + e^{l_z}), & S=X \\ \log(1 + e^{l_y}) - \log(e^{l_x} + e^{l_z}), & S=Y \\ \log(1 + e^{l_z}) - \log(e^{l_x} + e^{l_y}), & S=Z \end{cases}$$

③ At each iteration, have an estimation

$$\tilde{V}_i = \begin{cases} I(0), & \text{if } l_x, l_y, l_z < 0 \\ X(1), & \text{if } l_x > 0, l_x > l_y, l_z \\ Y(2), & \text{if } l_y > 0, l_y > l_x, l_z \\ Z(3), & \text{if } l_z > 0, l_z > l_x, l_y \end{cases}$$

④ Check Syndrome values

→ If all syndromes are 0. → Stop.

⑤ Check If decoder succeed or not.

→ If syndromes are all zero — success.

→ If Syndromes are not all zero.

→ Check if  $\tilde{V}_i$  belongs to the stabilizer group  
 ↑ CFor CSS codes:

① Have a systematic form of  $H_x, H_z$  at the beginning  
 $(\tilde{H}_x, \tilde{H}_z)$

② split  $\tilde{V}$  into  $X, Z$  errors  $V_x, V_z$ .

Use  $\tilde{H}_x, \tilde{H}_z$  to evaluate if  $V_x \in \text{Span}(\tilde{H}_x)$   
 $V_z \in \text{Span}(\tilde{H}_z)$   
 ↑ rowspace.

→ If yes, convergent error

→ If no, non-convergent error

## Degenerate Errors in QLDPC Codes.

### Recall Definition of Stabilizer Codes.

$$S = \langle S_1, \dots, S_r \rangle, S_i \in \{I, X, Y, Z\}^{\otimes n}$$

$$\mathcal{C} = \{ |\psi\rangle \mid S_i |\psi\rangle = |\psi\rangle, \forall i \in [r], |\psi\rangle \in \mathbb{C}^{2n} \}.$$

If error  $\overline{EES}$ .  $E|\psi\rangle - |\psi\rangle \Rightarrow E$  does not need to be corrected  
 $\hookrightarrow$  "Degenerate" Errors

BP Decoder are likely to perform worse in Quantum Codes due to degenerate errors.

### Exist Schemes

- ① Ordered Statistics Decoder (OSD)
  - ② Stabilizer Inactivation (SI)
- } both are post-processing algorithms after performing BP.

### Ordered Statistics Decoder

errors after BP

(large percent)  
non-convergent — Cannot be categorized but can be solved.

convergent — logical errors

likely  
(recall Maxwell Decoder in BEC)

$$w_i = |l_i| = \left| \log \frac{P(x_i=0)}{P(x_i=1)} \right| \rightarrow \text{reliability of estimation} \quad \hat{e}_I = \begin{cases} 0 & l_i > 0 \\ 1 & l_i < 0. \end{cases}$$

Order  $w_i$  of all VNs:

$$w_{r+1} \leq w_{r+2} \leq \dots \leq w_m, \quad I = [m]/J \quad \text{information set}$$

- ①  $r = \text{rk}(H)$ , find the first  $r$  indices  $J\{j_1, j_2, \dots, j_r\}$ , s.t.,  $\text{rk}(H|_J) = r$ .
- ② Change  $\hat{e}_J$  into values such  $H \hat{e}' = S$ . (convergent). (\*)

$\uparrow$  That is not necessarily sufficient for finding  $\hat{e}'$ .

OSD-0 algorithm ↪ Modification: relax (\*) by allowing modification over to OSD-w. the least reliable  $w$  VNs in  $\hat{e}_I$ .

OSD-w:

find the  $x \in \mathbb{F}_2^m$  s.t. if replacing  $(\hat{e}_I)_{[w]}$  by  $x$ , the result  $\hat{e}'(x)$  obtained from ② is of the smallest weight (among all choices of  $x$ ).

## OSD in Quantum Codes

$$p_i \leftarrow \max \{ p_{i,x}, p_{i,y}, p_{i,z}, p_{i,\bar{z}} \}.$$

Order the Qubits by

$$p_{\sigma(1)} \leq p_{\sigma(2)} \leq \dots \leq p_{\sigma(n)}.$$

Use the binary representation to perform OSD-w algorithm.

$$H' = (H_x | H_z), \quad S' = (S_x | S_z).$$

$$\text{Obtain } \hat{e} = (\hat{e}_x | \hat{e}_z) \Rightarrow \tilde{e} = \hat{e}_x x + \hat{e}_z z.$$

## Stabilizer Inactivation

Stabilizer splitting errors:

X error  $e_x$ , row  $r_x$  in  $H_x$ .

$$\frac{w(e_x + r_x)}{n} = w(e_x^*) \rightarrow \text{hard to BP can go into either choice}$$

(splits the probability of the coset into smaller probabilities)

$$\mathcal{J}(e_x, e_x^*) = \sum_{i \in \text{supp}(e_x)} |\tilde{p}_i| - \sum_{i \in \text{supp}(e_x^*)} |\tilde{p}_i| = \sum_{i \in \text{supp}(r_x)} (-1)^{\text{rep}} |\tilde{p}_i| \quad \text{of different representatives),}$$

$$\leq \sum_{i \in \text{supp}(r_x)} |\tilde{p}_i| = f(r_x).$$

Small  $f(r_x)$  is more likely to lead to small  $\delta(x, x')$ 's, and thus is more likely to be unreliable.

Inactivate least reliable stabilizer  $R_x$ :

$$H\hat{z}^* = \left[ \begin{array}{c|c} H\hat{z}|_{r_x} & A \\ \hline & H\hat{z}|_{\bar{r}_x} \end{array} \right] \quad \text{use } H\hat{z}|_{\bar{r}_x} \text{ to redo BP to obtain } \hat{e}_x|r_x$$

Scan through  $r_{x,\sigma(i)}$  with  $f(r_{x,\sigma(0)}) \leq f(r_{x,\sigma(1)}) \leq \dots \leq f(r_{x,\sigma(n)})$

until  $H\hat{z}|_{\bar{r}_x} \hat{e}_z|\bar{r}_x = S_x|\bar{r}_x$ , and  $H\hat{z}|_{r_x} \hat{e}_x|r_x = S_x|r_x + A\hat{e}_x|r_x$  has a solution.