

Improved Synthesis of Toffoli-Hadamard Circuits

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Quantum Algorithm: Carrying out a well-defined task.

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Quantum Compilation: Program \Rightarrow Sequence of elementary quantum gates.

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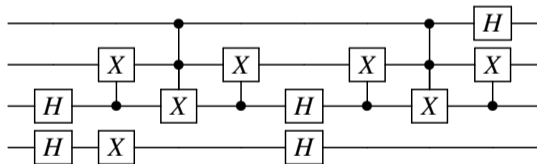
Quantum Compilation: Program \Rightarrow Sequence of elementary quantum gates.

Implementation: Mapping unitary operations to physical architectures.

Restricted Clifford+T Circuits¹

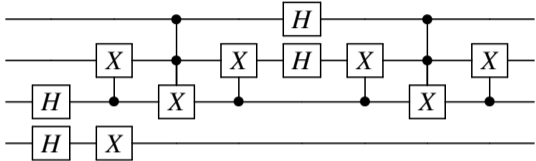
Toffoli-Hadamard circuits are quantum circuits over the gate set

$$\{X, CX, CCX, H\}.$$

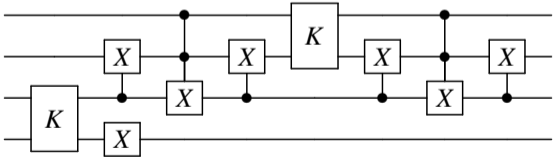


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Restricted Clifford+T Circuits



A Toffoli-Hadamard Circuit



A Toffoli-K Circuit

Basic Gates

$$(-1) = [-1]$$

$$H = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, \quad K = H \otimes H = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad CX = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \left[\begin{array}{c|c} I_2 & \mathbf{0} \\ \mathbf{0} & X \end{array} \right], \quad CCX = \left[\begin{array}{c|c} I_6 & \mathbf{0} \\ \mathbf{0} & X \end{array} \right]$$

The Circuit-Matrix Correspondence

- A family of quantum circuits corresponds to a group of matrices.
- Studying matrix groups is a way to study quantum circuits.

Orthogonal Scaled Dyadic Matrices

- \mathcal{L}_n is the group of **orthogonal scaled dyadic matrices**, which consists of $n \times n$ orthogonal matrices of the form $M/\sqrt{2}^k$, where M is an integer matrix and k is a nonnegative integer.

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- Example: $V \in \mathcal{L}_4$

$$V = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & -1 & 0 & -1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$$

Orthogonal Dyadic Matrices

- $\mathbb{Z}\left[\frac{1}{2}\right] = \left\{\frac{u}{2^q} \mid u \in \mathbb{Z}, q \in \mathbb{N}\right\}$ is the ring of *dyadic fractions*.

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- Example: $U \in O_5$

$$U = \begin{bmatrix} 3/4 & 1/4 & -1/4 & 1/4 & 1/2 \\ 1/4 & 3/4 & 1/4 & -1/4 & -1/2 \\ -1/4 & 1/4 & 3/4 & 1/4 & 1/2 \\ 1/4 & -1/4 & 1/4 & 3/4 & -1/2 \\ -1/2 & 1/2 & -1/2 & 1/2 & 0 \end{bmatrix}$$

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- $O_n \subset \mathcal{L}_n$.

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Constructive Membership Problem (CMP)

Let \mathcal{G} be a group and let S be a set of generators for \mathcal{G} . The *constructive membership problem for \mathcal{G} and S* , denoted $\mathcal{P}(\mathcal{G}, S)$, is the following:

Given $g \in \mathcal{G}$, find a sequence of generators $s_1, \dots, s_\ell \in S$ such that

$$s_1 \cdot \dots \cdot s_\ell = g,$$

where \cdot is the group operation.

- The smaller the ℓ , the better the solution.

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- The smaller the ℓ , the better the solution.
- A solution is **optimal** if the sequence is a **shortest possible sequence**.
- An algorithm to solve the CMP is called an **exact synthesis algorithm**.

The Circuit-Matrix Correspondence I

Theorem (Solutions to CMP: The AGR Algorithm¹)

For an n -dimensional orthogonal matrix U ,

- it can be exactly represented by a circuit over $\{X, CX, CCX, H\}$ iff $U \in \mathcal{L}_n$.
- it can be exactly represented by a circuit over $\{X, CX, CCX, K\}$ iff $U \in \mathcal{O}_n$.

The gate complexity of the AGR algorithm in both cases is $O(2^n \log(n)k)$.

- A good solution to CMP yields shorter quantum circuits.

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- A good solution to CMP yields shorter quantum circuits.
- **Can we find a good solution to the CMP for \mathcal{O}_n and \mathcal{L}_n ?**

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Two-level Operator: $U_{[\alpha,\beta]}$

Definition

Let $U = \begin{bmatrix} x_{1,1} & x_{1,2} \\ x_{2,1} & x_{2,2} \end{bmatrix}$. The action of $U_{[\alpha,\beta]}$, $1 \leq \alpha < \beta \leq n$, is defined as

$$U_{[\alpha,\beta]}v = w, \text{ where } \begin{cases} \begin{bmatrix} w_\alpha \\ w_\beta \end{bmatrix} = U \begin{bmatrix} v_\alpha \\ v_\beta \end{bmatrix}, \\ w_i = v_i, i \notin \{\alpha, \beta\}. \end{cases}$$

Example:

$$\text{Let } X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}. \text{ Then } X_{[2,3]} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \text{ and } X_{[2,3]} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{bmatrix} = \begin{bmatrix} v_1 \\ v_3 \\ v_2 \\ v_4 \end{bmatrix}.$$

Four-level Operator: $U_{[\alpha,\beta,\gamma,\delta]}$

Similarly, we can create a four-level operator by embedding a 4×4 matrix U into an $n \times n$ identity matrix.

$$\text{Let } K = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}. \text{ Then } K_{[1,2,4,6]} = \begin{bmatrix} 1/2 & 1/2 & 0 & 1/2 & 0 & 1/2 \\ 1/2 & -1/2 & 0 & 1/2 & 0 & -1/2 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1/2 & 1/2 & 0 & -1/2 & 0 & -1/2 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 1/2 & -1/2 & 0 & -1/2 & 0 & 1/2 \end{bmatrix}.$$

$$K_{[1,2,4,6]} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \end{bmatrix} = \begin{bmatrix} (v_1 + v_2 + v_4 + v_6)/2 \\ (v_1 - v_2 + v_4 - v_6)/2 \\ v_3 \\ (v_1 + v_2 - v_4 - v_6)/2 \\ v_5 \\ (v_1 - v_2 - v_4 + v_6)/2 \end{bmatrix}.$$

The Circuit-Matrix Correspondence II

$$\mathcal{F}_n = \{(-1)_{[\alpha]}, X_{[\alpha,\beta]}, K_{[\alpha,\beta,\gamma,\delta]}, I_{n/2} \otimes H : 1 \leq \alpha < \beta < \gamma < \delta \leq n\}.$$

$$\mathcal{G}_n = \{(-1)_{[\alpha]}, X_{[\alpha,\beta]}, K_{[\alpha,\beta,\gamma,\delta]} : 1 \leq \alpha < \beta < \gamma < \delta \leq n\}.$$

Theorem (Solutions to CMP: The AGR Algorithm¹)

Let U be an $n \times n$ matrix.

- $U \in \mathcal{L}_n$ iff U can be written as a product of elements of \mathcal{F}_n .
- $U \in \mathcal{O}_n$ iff U can be written as a product of elements of \mathcal{G}_n .

- When $n = 2^m$, every operator in \mathcal{G}_n and \mathcal{F}_n can be exactly represented by $O(\log(n))$ operators in $\{X, CX, CCX, K\}$ and $\{X, CX, CCX, H\}$, respectively.

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The Least Denominator Exponent (LDE)

Let $t \in \mathbb{Z}[\frac{1}{2}]$. $t = \frac{a}{2^k}$, where $a \in \mathbb{Z}$ and $k \in \mathbb{N}$. k is a *denominator exponent* for t . The minimal such k is called the **least denominator exponent** of t , written $\text{lde}(t)$.

Example: LDE of a column vector

$$v = \frac{1}{2^7} \begin{bmatrix} 54 \\ 62 \\ 98 \\ 2 \\ 2 \\ 2 \\ 2 \\ 2 \end{bmatrix} = \frac{2}{2^7} \begin{bmatrix} 27 \\ 31 \\ 49 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{2^6} \begin{bmatrix} 27 \\ 31 \\ 49 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\text{lde}(v) = 6$$

Example: LDE of a matrix

$$U = \frac{1}{2} \begin{bmatrix} -1 & 1 & 1 & 0 & 1 & 0 & 0 & 0 \\ -1 & -1 & 0 & 1 & 0 & 1 & 0 & 0 \\ -1 & 1 & -1 & 0 & -1 & 0 & 0 & 0 \\ -1 & -1 & 0 & -1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 & -1 & -1 & 0 & 0 \\ 0 & 0 & 1 & -1 & -1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 \end{bmatrix}$$

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Lemma (Base Case)

Let $v \in \mathbb{Z}\left[\frac{1}{2}\right]^n$ be a unit vector. Let $k = \text{lde}(v)$. If $k = 0$, then $v = \pm e_j$ for some $j \in \{1, \dots, n\}$.

Lemma (Count)

Let $v \in \mathbb{Z}\left[\frac{1}{2}\right]^n$ be a unit vector, and $\text{lde}(v) = k > 0$. Let $w = 2^k v$. Then the number of odd entries in w is a multiple of 4.

Lemma (Parity Reduction)

Let u_1, u_2, u_3, u_4 be odd integers. Then there exist $\tau_1, \tau_2, \tau_3, \tau_4 \in \mathbb{Z}_2$ such that

$$K_{[1,2,3,4]} (-1)_{[1]}^{\tau_1} (-1)_{[2]}^{\tau_2} (-1)_{[3]}^{\tau_3} (-1)_{[4]}^{\tau_4} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \end{bmatrix} = \begin{bmatrix} u'_1 \\ u'_2 \\ u'_3 \\ u'_4 \end{bmatrix}, \quad u'_1, u'_2, u'_3, u'_4 \text{ are even integers.}$$

The AGR Algorithm (II)

- The algorithm proceeds one column at a time, reducing each column to a corresponding basis vector.
- While outputting a word \vec{G}_ℓ after each iteration, the algorithm recursively acts on the input matrix until it is reduced to the identity matrix \mathbb{I} .


$$M \xrightarrow{\vec{G}_1} \left(\begin{array}{ccc|c} & & & 0 \\ & M' & & \vdots \\ & & & 0 \\ \hline 0 & \dots & 0 & 1 \end{array} \right) \xrightarrow{\vec{G}_2} \left(\begin{array}{ccc|cc} & & & 0 & 0 \\ & M'' & & \vdots & \vdots \\ & & & 0 & 0 \\ \hline 0 & \dots & 0 & 1 & 0 \\ 0 & \dots & 0 & 0 & 1 \end{array} \right) \xrightarrow{\vec{G}_3} \dots \xrightarrow{\vec{G}_\ell} \mathbb{I}$$

$$\vec{G}_\ell \cdots \vec{G}_1 M = \mathbb{I} \Rightarrow M = \vec{G}_1^{-1} \cdots \vec{G}_\ell^{-1}$$

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\mathbf{e}_n
 \mathbf{e}_{n-1}

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Gate Complexity of the AGR Algorithm

Lemma

Let $\mathbf{u} \in \mathbb{Z}\left[\frac{1}{2}\right]^n$ with $\text{lde}(\mathbf{u}) = k$. The number of generators in \mathcal{G}_n to reduce \mathbf{u} to \mathbf{e}_j is $O(nk)$.

Theorem

Let $U \in O_n$ with $\text{lde}(U) = k$. U can be exactly represented by $O(2^n k)$ generators over \mathcal{G}_n .

Proof. Let $f_{\mathbf{u}_i}$ be the cost of reducing \mathbf{u}_i to \mathbf{e}_i .

- Each row operation may increase the lde of any column in U by 1.
- During reduction, the lde of any other column may increase up to $2k$.

$$f_{\mathbf{u}_1} = O(nk), \quad f_{\mathbf{u}_2} = O((n-1)2k), \quad f_{\mathbf{u}_3} = O((n-2)2^2k), \quad \dots, \quad f_{\mathbf{u}_n} = O(2^{n-1}k).$$

$$S_n = \sum_{i=1}^n f_{\mathbf{u}_i} = \sum_{i=1}^n (n-i+1)2^{i-1}k = O(2^n k).$$

The Householder Algorithm²

With **ancillary qubits**, the gate complexity of the exact synthesis for \mathcal{L}_n over \mathcal{F}_n is reduced from $O(2^nk)$ to $O(n^2k)$.

²Vadym Kliuchnikov (2013). "Synthesis of unitaries with Clifford+ T circuits". In: *arXiv preprint arXiv:1306.3200*.

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Definition

For an n -dimensional unit vector $|\psi\rangle$, the reflection operator around $|\psi\rangle$ is

$$R_{|\psi\rangle} = I - 2|\psi\rangle\langle\psi|.$$

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Proposition: Gate Complexity of the Reflection Operator

Let $|\psi\rangle = \mathbf{v}/\sqrt{2^k}$ be an n -dimensional unit vector with $\text{Ide}_{\sqrt{2}}(|\psi\rangle) = k$, \mathbf{v} is an integer vector. The reflection operator $R_{|\psi\rangle}$ can be exactly represented by $O(nk)$ generators over \mathcal{F}_n .

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Unitary Simulation

Let $U \in \mathcal{L}_n$. Then U can be simulated using the unitary

$$U' = |+\rangle\langle -| \otimes U + |-\rangle\langle +| \otimes U^\dagger.$$

Moreover, $U' \in \mathcal{L}_{2n}$ and U' can be factored as a product $U' = \prod_{j=1}^n R_{|\omega_j^-\rangle}$ of reflection operators around vectors

$$|\omega_j^-\rangle = \frac{(|-\rangle|j\rangle - |+\rangle|\mathbf{u}_j\rangle)}{\sqrt{2}},$$

\mathbf{u}_j is the j -th column vector in U and $|j\rangle$ is the j -th computational basis vector.

Unitary Simulation

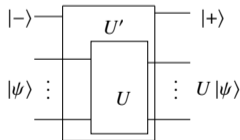
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Moreover, $U' \in \mathcal{L}_{2n}$ and U' can be factored as a product $U' = \prod_{j=1}^n R_{|\omega_j^-\rangle}$ of reflection operators around vectors

$$|\omega_j^-\rangle = \frac{(|-\rangle|j\rangle - |+\rangle|u_j\rangle)}{\sqrt{2}},$$

u_j is the j -th column vector in U and $|j\rangle$ is the j -th computational basis vector.



Unitary Simulation

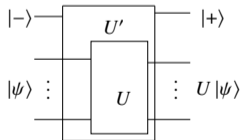
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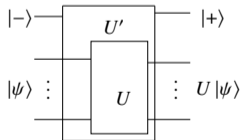
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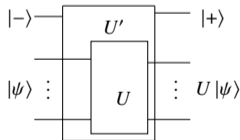
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$$I - U' = 2 \sum_{j=1}^n |\omega_j^-\rangle\langle\omega_j^-| \Rightarrow U' = I - 2 \sum_{j=1}^n |\omega_j^-\rangle\langle\omega_j^-| = \prod_{j=1}^n R_{|\omega_j^-\rangle}.$$

Gate Complexity of the Householder Algorithm

Theorem

Let $U \in \mathcal{L}_n$ with $\text{lde}_{\sqrt{2}}(U) = k$. Then U can be represented by $O(n^2k)$ generators over \mathcal{F}_n using the Householder algorithm.

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Proof. We showed that U can be simulated by U' where

$$U' = |+\rangle\langle -| \otimes U + |- \rangle\langle +| \otimes U^\dagger = \prod_{j=1}^n R_{|\omega_j^-\rangle}.$$

Moreover, each $R_{|\omega_j^-\rangle}$ can be exactly represented by $O(nk)$ generators from \mathcal{F}_n .

Therefore, to represent U , we need $n \cdot O(nk) = O(n^2k)$ generators over \mathcal{F}_n . \square

The Global Synthesis Algorithm

- The AGR algorithm carries out matrix factorization **locally** - it synthesizes one column at a time.

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- Define a global synthesis method for $U \in \mathcal{L}_8$, then **leverage** this to find a global synthesis method for $U' \in \mathcal{O}_8$.

Intuition

$U \in \mathcal{L}_8$. Write $U = \frac{1}{\sqrt{2}^k} M$ with k minimal. There exists $\vec{G}_1, \dots, \vec{G}_k$ over \mathcal{F} , such that

$$\frac{1}{\sqrt{2}^k} M \xrightarrow{\vec{G}_1} \frac{1}{\sqrt{2}^{k-1}} M' \xrightarrow{\vec{G}_2} \frac{1}{\sqrt{2}^{k-2}} M'' \xrightarrow{\vec{G}_3} \dots \xrightarrow{\vec{G}_k} \mathbb{I}.$$

Therefore,

$$\vec{G}_k \cdots \vec{G}_1 U = \mathbb{I} \implies U = \vec{G}_1^{-1} \cdots \vec{G}_k^{-1}.$$

Binary Pattern

Let $U \in \mathcal{L}_n$. Write $U = \frac{1}{\sqrt{2}^k} M$ with k minimal. The residue mod 2 of M is called the **binary pattern** of U , denoted as \bar{U} .

Example: $U \in \mathcal{L}_5$

$$U = \frac{1}{\sqrt{2}^4} \begin{bmatrix} 3 & 1 & -1 & 1 & 2 \\ 1 & 3 & 1 & -1 & -2 \\ -1 & 1 & 3 & 1 & 2 \\ 1 & -1 & 1 & 3 & -2 \\ -2 & 2 & -2 & 2 & 0 \end{bmatrix} \rightarrow \bar{U} = \begin{bmatrix} 1 & 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Number Theoretic Property I

Weight Lemma

Let $U \in \mathcal{L}_n$ and $\text{lde}_{\sqrt{2}}(U) = k \geq 2$. Let \mathbf{u} be an arbitrary column vector in \overline{U} . Then

$$|\{u_i; u_i = 1, 1 \leq i \leq n\}| \equiv 0(4).$$

In other words, in each column of \overline{U} , the 1's occur in quadruples.

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Proof. Let \mathbf{v} be a column vector in U and $\mathbf{v} = \frac{1}{\sqrt{2}^k} \mathbf{w}$, where $\mathbf{w} \in \mathbb{Z}^n$. Since $\langle \mathbf{v}, \mathbf{v} \rangle = 1$, $\langle \mathbf{w}, \mathbf{w} \rangle = 2^k$ and thus $\sum w_i^2 = 2^k$. When $k \geq 2$, $\sum w_i^2 \equiv 0(4)$. Note that

$$w_i^2 \equiv 1(4) \iff w_i \equiv 1(2), \quad w_i^2 \equiv 0(4) \iff w_i \equiv 0(2).$$

Hence the number of odd entries in \mathbf{w} is a multiple of 4. □

Number Theoretic Property II

Intuition: The 1's in any two distinct columns of \bar{U} collide evenly many times.

Collision Lemma

Let $U \in \mathcal{L}_n$ and $\text{lde}_{\sqrt{2}}(U) = k > 0$. Any two distinct columns in \bar{U} must have evenly many 1's in common.

Example: Evenly many collisions

$$u_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

Example: Oddly many collisions

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Binary Patterns of \mathcal{L}_8

Theorem

There exists a set \mathcal{P} of 14 binary patterns such that if $U \in \mathcal{L}_8$ and $\text{lde}(U) \geq 2$, then $\bar{U} \in \mathcal{P}$ (up to row and column permutations, as well as taking transpose).

Proof. By a long case distinction using the **Weight** and **Collision Lemmas**.

Definition

A matrix $\bar{U} \in \mathbb{Z}_2^{8 \times 8}$ is **row-paired** if identical rows occur evenly many times.

Definition

A matrix $\bar{U} \in \mathbb{Z}_2^{8 \times 8}$ is **column-paired** if identical columns occur evenly many times.

Remark: We demonstrate an example and a counterexample when $n = 4$.

Example: Row-paired and column-paired

$$\bar{U} = \begin{bmatrix} 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

Example: Only column-paired

$$\bar{V} = \begin{bmatrix} 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 \end{bmatrix}$$

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When the Binary Pattern is “Nice”

Theorem (Row-paired Reduction)

If $U \in \mathcal{L}_8$ and \bar{U} is row-paired, then there exists $P \in S_8$ such that $\text{lde}_{\sqrt{2}}(((I \otimes H) P)U) < \text{lde}_{\sqrt{2}}(U)$.

Theorem (Column-paired Reduction)

If $U \in \mathcal{L}_8$ and \bar{U} is column-paired, then there exists $P \in S_8$ such that $\text{lde}_{\sqrt{2}}(U(P(I \otimes H))) < \text{lde}_{\sqrt{2}}(U)$.

Remark: Below we sketch the proof for the **Row-paired Reduction** using a 6×6 matrix as an example.

Proof. Consider $U \in \mathcal{L}_6$ with $\text{lde}_{\sqrt{2}}(U) = k$. Since \bar{U} is row-paired, there exists $P \in S_6$ such that

$$PU = \frac{1}{\sqrt{2}^k} \begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_6 \end{bmatrix}, \text{ where } r_1 \equiv r_2(2), r_3 \equiv r_4(2), r_5 \equiv r_6(2). \text{ Now}$$

$$H = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \text{ and } I \otimes H = \left[\begin{array}{c|c|c} H & \mathbf{0} & \mathbf{0} \\ \hline \mathbf{0} & H & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{0} & H \end{array} \right]. \text{ Therefore,}$$

$$(I \otimes H)PU = \frac{1}{\sqrt{2}^{k+1}} \begin{bmatrix} r_1 + r_2 \\ r_1 - r_2 \\ r_3 + r_4 \\ r_3 - r_4 \\ r_5 + r_6 \\ r_5 - r_6 \end{bmatrix} = \frac{2}{\sqrt{2}^{k+1}} \begin{bmatrix} r'_1 \\ \vdots \\ r'_6 \end{bmatrix} = \frac{1}{\sqrt{2}^{k-1}} \begin{bmatrix} r'_1 \\ \vdots \\ r'_6 \end{bmatrix}, \text{ where } r'_1, \dots, r'_6 \in \mathbb{Z}^{1 \times 6}.$$

Hence $\text{lde}_{\sqrt{2}}((I \otimes H)PU) < \text{lde}_{\sqrt{2}}(U)$, for some $P \in S_6$.

When the Binary Pattern is “NOT Nice”

Theorem

Consider $U \in \mathcal{L}_8$ and \bar{U} is neither row-paired nor column-paired. Let $U' = (I \otimes H) U (I \otimes H)$. Then \bar{U}' is row-paired and $\text{lde}_{\sqrt{2}}(U') \leq \text{lde}_{\sqrt{2}}(U)$.

Proof. By direct computation.

Theorem

Let $U \in \mathcal{L}_8$ and $\text{lde}_{\sqrt{2}}(U) = k$. Then there exists C over \mathcal{F} such that $[[C]] = U$ and the length of C is $O(k)$.

Proof. Let $U \in \mathcal{L}_8$, proceed by induction on k .

- $k \leq 1$, there exists C composed of $(-1)_{[\alpha]}$, $X_{[\alpha,\beta]}$ and $I \otimes H$ such that $[[C]] = U$ and the length of C is $O(1)$.
- $k \geq 2$, \bar{U} must be one of the 14 binary patterns.
 - * If \bar{U} is nice, then $\text{lde}((I \otimes H)PU) \leq k - 1$ and proceed recursively with $(I \otimes H)PU$.
 - * If \bar{U} is not nice, then $(I \otimes H)U(I \otimes H)$ is nice so $\text{lde}((I \otimes H)P(I \otimes H)U(I \otimes H)) \leq k - 1$ and proceed recursively with $(I \otimes H)P(I \otimes H)U(I \otimes H)$.

Generator Relations for \mathcal{L}_8 and O_8^3

- \mathcal{L}_8 is generated by $\mathcal{F} = \{(-1)_{[\alpha]}, X_{[\alpha,\beta]}, K_{[\alpha,\beta,\gamma,\delta]}, I \otimes H : 1 \leq \alpha < \beta < \gamma < \delta \leq 8\}$.
- O_8 is generated by $\mathcal{G} = \{(-1)_{[\alpha]}, X_{[\alpha,\beta]}, K_{[\alpha,\beta,\gamma,\delta]} : 1 \leq \alpha < \beta < \gamma < \delta \leq 8\}$.

$$(I \otimes H)(I \otimes H) = \epsilon \tag{1}$$

$$(I \otimes H)(-1)_{[1]} = (-1)_{[1]}X_{[1,2]}(-1)_{[1]}(I \otimes H) \tag{2}$$

$$(I \otimes H)X_{[a,a+1]} = (-1)_{[a+1]}^{a+1}X_{[a,a+1]}^a K_{[a-1,a,a+1,a+2]}^a (I \otimes H) \tag{3}$$

$$(I \otimes H)K_{[1,2,3,4]} = K_{[1,2,3,4]}(I \otimes H) \tag{4}$$

Intuition: Commuting $I \otimes H$ with an element in \mathcal{G} adds $O(1)$ gates.

³Sarah Meng Li, Neil J Ross, and Peter Selinger (2021). “Generators and relations for the group $O_n(\mathbb{Z}[1/2])$ ”. In: *arXiv preprint arXiv:2106.01175*.

Lemma

For any M over \mathcal{G} , there exists M' over \mathcal{G} such that $(I \otimes H) M = M' (I \otimes H)$.
Moreover, if M has length $O(m)$, then M' has length $O(m)$.

Example:

$$\begin{aligned}(I \otimes H)K_{[1,2,3,4]}(-1)_{[1]}X_{[1,2]}(I \otimes H) &= K_{[1,2,3,4]}(I \otimes H)(-1)_{[1]}X_{[1,2]}(I \otimes H) \\ &= K_{[1,2,3,4]}(-1)_{[1]}X_{[1,2]}(-1)_{[1]}(I \otimes H)X_{[1,2]}(I \otimes H) \\ &= K_{[1,2,3,4]}(-1)_{[1]}X_{[1,2]}(-1)_{[1]}(-1)_{[2]}(I \otimes H)(I \otimes H) \\ &= K_{[1,2,3,4]}(-1)_{[1]}X_{[1,2]}(-1)_{[1]}(-1)_{[2]}.\end{aligned}$$

Theorem

Let $U \in \mathcal{O}_8$ and $\text{lde}(U) = k \geq 1$. Then there exists C over \mathcal{G} such that $[[C]] = U$ and the length of C is $O(k)$.

Proof. Let $U \in \mathcal{O}_8$ and $\text{lde}(U) = k$. Then $U \in \mathcal{L}_8$ with $\text{lde}_{\sqrt{2}}(U) = 2k$. Using the global synthesis for \mathcal{L}_8 , we can express U as a word W over \mathcal{F} with evenly many occurrences of $I \otimes H$, and the length of W is $O(k)$. Consider any subword W_i of the form

$$(I \otimes H) C (I \otimes H),$$

where C does not contain $I \otimes H$.

Theorem

Consider $U \in O_8$ and $\text{lde}(U) = k \geq 1$. Then there exists C over \mathcal{G} such that $[[C]] = U$ and the length of C is $O(k)$.

Proof Continued. Suppose the length of W_i is $O(k)$. Then

$$W_i = (I \otimes H) C (I \otimes H) \rightarrow C' (I \otimes H) (I \otimes H) \rightarrow C'$$

W_i can be rewritten as a word C' over \mathcal{G} of length at most $3 * O(k)$ generators. Hence we can rewrite W as a word W' over \mathcal{G} of length no more than $3 * O(k)$.

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- **Extend** the global synthesis to arbitrary dimensions: O_n and \mathcal{L}_n .
- **Present** the global synthesis results of O_n and \mathcal{L}_n in terms of the restricted Clifford+T circuits over $\{X, CX, CCX, K\}$ and $\{X, CX, CCX, H\}$ respectively.

A scenic view of a beach at sunset. The sun is low on the horizon, casting a warm, golden glow over the scene. The sky is filled with wispy, white clouds. The ocean is a deep blue, with waves breaking onto the shore, creating white foam. The sand in the foreground is a rich, golden-brown color, reflecting the light from the sun. The overall atmosphere is peaceful and serene.

Thank you!