

QFlux: An Open-Source Python Package for Quantum Dynamics Simulations

QFlux: Quantum Circuit Implementations of Molecular Dynamics

Victor S Batista

Yale University, Department of Chemistry and Yale Quantum Institute

Part III: The QFlux Synthesis Pipeline

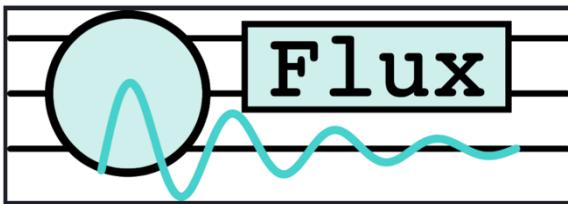
*Every quantum algorithm depends on two operations:
preparing a state and implementing a unitary.*

These are not physical problems — they're synthesis problems.

How linear algebra is compiled as pulses applied to quantum hardware

<https://qflux.batistalab.com>

[Part_III.ipynb](#)



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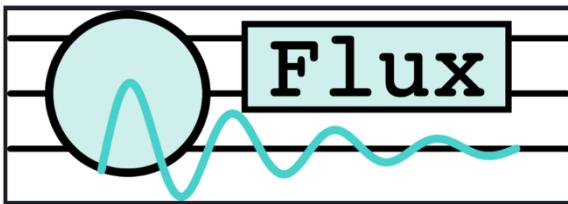
This tutorial is based on the manuscript

QFlux: Quantum Circuit Implementations for Molecular Dynamics

Part III – State Initialization and Unitary Decomposition

Authors:

Alexander V. Soudackov, Delmar G. A. Cabral, Brandon C. Allen, Xiaohan Dan, Nam P. Vu, Cameron Cianci, Rishab Dutta, Sabre Kais, Eitan Geva, and Victor S. Batista



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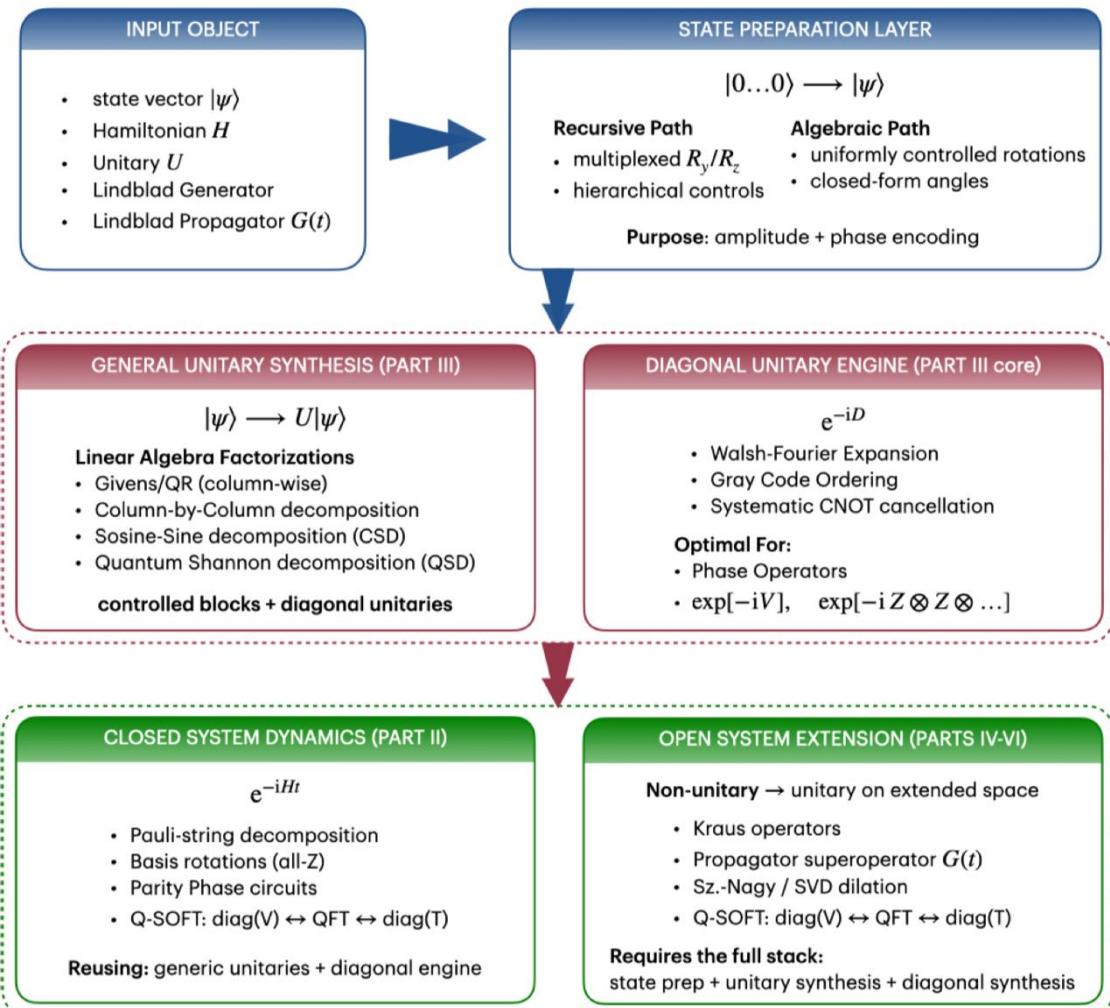
The QFlux Synthesis Pipeline

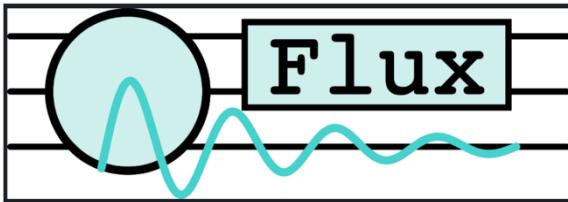
The backbone of QFlux

One pipeline for all simulations

- State vectors → state preparation
- Generic unitaries → linear-algebra factorizations
- Diagonal operators → Walsh synthesis

Reused across closed and open systems





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Method Selection: Which Tool When?

State preparation:

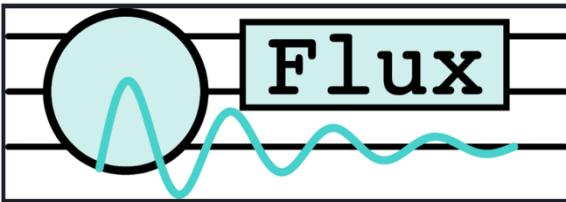
- Multiplexors → transparent, good for debugging
- UCRs → compact, compiler-friendly

Unitaries:

- Givens / QR → pedagogical baseline
- Column-by-column → practical
- CSD / QSD → asymptotically optimal

Diagonal:

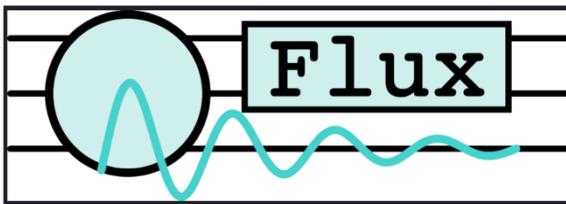
- Walsh + Gray code → NISQ-optimal



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PART A — STATE PREPARATION

- **Input:** $|00 \dots 0\rangle = |0\rangle^{\otimes n} = \underbrace{\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \otimes \dots \otimes \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}}_{n \text{ qubits}} = \boxed{(1, 0, \dots, 0)^\top}$
- **Target:** $|\psi\rangle = \sum_{k_1, k_2, \dots, k_n \in \{0,1\}} c_{k_1 k_2 \dots k_n} |k_1 k_2 \dots k_n\rangle$
 $|\psi\rangle = \boxed{(c_1, c_2, \dots, c_{2^n})^\top}$
- **Goal:** deterministic, exact preparation
- **Strategy:** *disentangle qubits one by one*

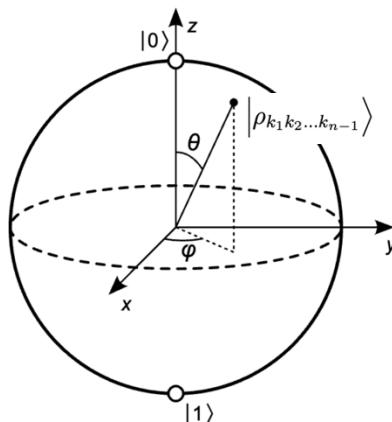


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PART A — STATE PREPARATION

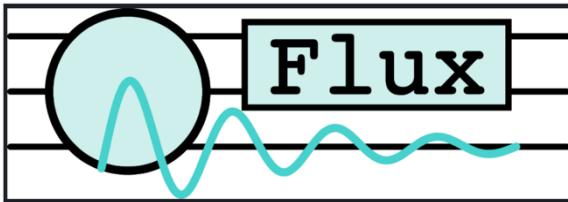
$$|\psi\rangle = \sum_{k_1, k_2, \dots, k_n \in \{0,1\}} c_{k_1 k_2 \dots k_n} |k_1 k_2 \dots k_n\rangle$$

$$|\psi\rangle = \sum_{k_1, k_2, \dots, k_{n-1} \in \{0,1\}} |k_1 k_2 \dots k_{n-1}\rangle \otimes \underbrace{\left[c_{k_1 k_2 \dots k_{n-1} 0} |0\rangle + c_{k_1 k_2 \dots k_{n-1} 1} |1\rangle \right]}_{\rho_{k_1 k_2 \dots k_{n-1}}}$$



$$|\rho_{k_1 k_2 \dots k_{n-1}}\rangle = c_{k_1 k_2 \dots k_{n-1} 0} |0\rangle + c_{k_1 k_2 \dots k_{n-1} 1} |1\rangle$$

$$R_y(-\theta_{k_1 k_2 \dots k_{n-1}}) R_z(-\varphi_{k_1 k_2 \dots k_{n-1}}) |\rho_{k_1 k_2 \dots k_{n-1}}\rangle = r_{k_1 k_2 \dots k_{n-1}} e^{i t_{k_1 k_2 \dots k_{n-1}}} |0\rangle$$



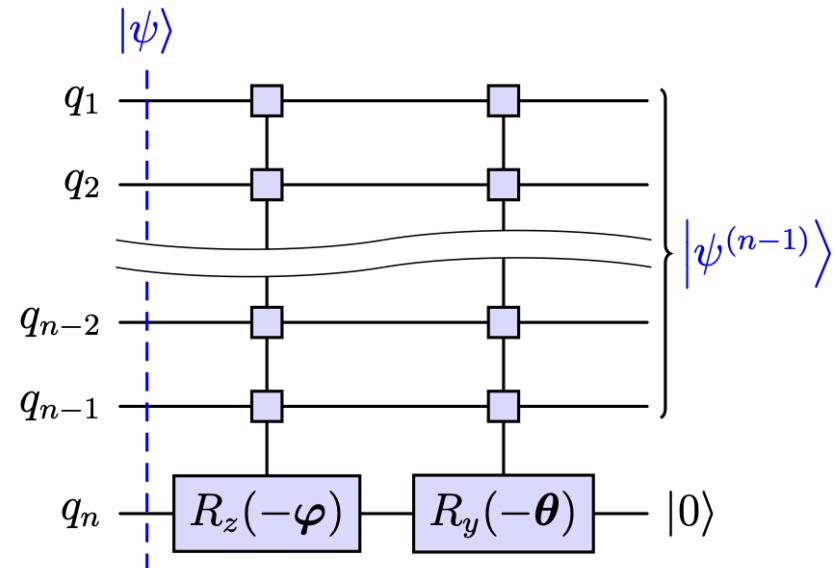
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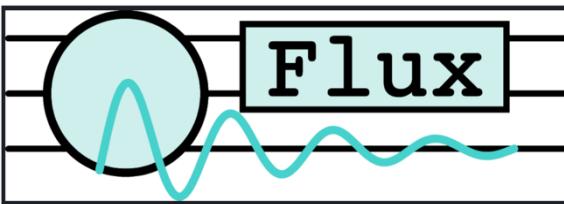
Multiplexor Intuition

$$R^{(n)} = \bigoplus_{k_1, \dots, k_{n-1} \in \{0,1\}} R_y(-\theta_{k_1 k_2 \dots k_{n-1}}) R_z(-\varphi_{k_1 k_2 \dots k_{n-1}})$$

$$R^{(n)} |\psi\rangle = |\psi^{(n-1)}\rangle \otimes |0\rangle$$

- Group amplitudes by last qubit
- Each conditional single-qubit state lies on Bloch sphere
- Rotate each to $|0\rangle$ using conditional rotations
- Remove entanglement one qubit at a time





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What Is a Multiplexor?

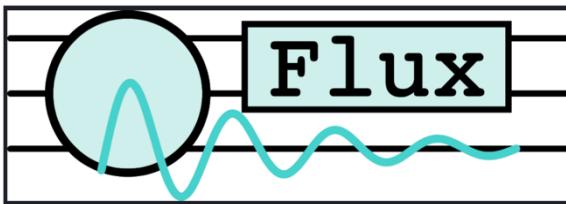
For each bitstring of the control qubits, apply a different Ry–Rz pair to the target

- A gate that applies different single-qubit rotations depending on controls
- Block-diagonal structure:
 - One rotation per control configuration
- Acts as “conditional disentangler”

▶

```
1 from scipy.linalg import block_diag
2
3 def multiplexor_matrix(n, vector, bit=0):
4     bit = int(bool(bit))
5     multiplexor = None
6     for i in np.arange(0,2**n,2):
7         c0, c1 = vector[i], vector[i+1]
8         theta, phi = compute_bloch_angles(c0, c1)
9         r = ry_matrix(bit*np.pi - theta) @ rz_matrix(-phi)
10        multiplexor = block_diag(multiplexor, r) if multiplexor is not None else r
11
12 return multiplexor
```

[Script S.1.3: Part_III.ipynb](#)



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Recursive Disentanglement Algorithm

- Step 1: rotate last qubit $\rightarrow |0\rangle$
- Step 2: recurse on remaining $n-1$ qubits
- Reverse the sequence to prepare $|\psi\rangle$
- Cost: $\sim 2^n$ CNOTs

Script S.1.5: [Part_III.ipynb](#)

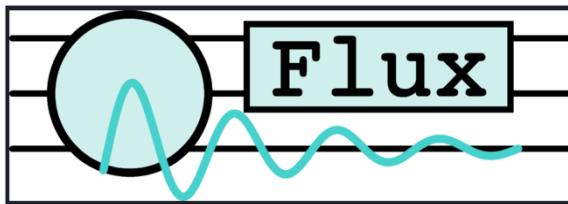
Example: Three-Qubit State Preparation

Script S.1.6: [Part_III.ipynb](#)

Example: Coherent Wavepacket State Preparation on a 6-Qubit System

Script S.1.5: Example: Three-Qubit State Preparation

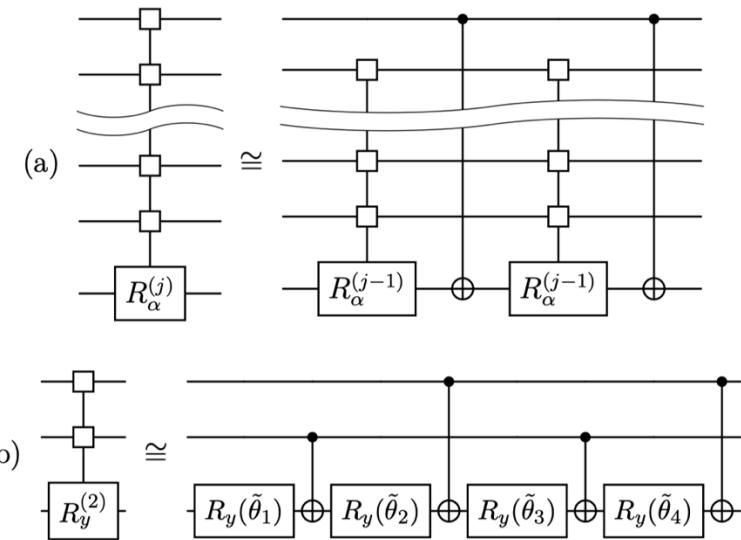
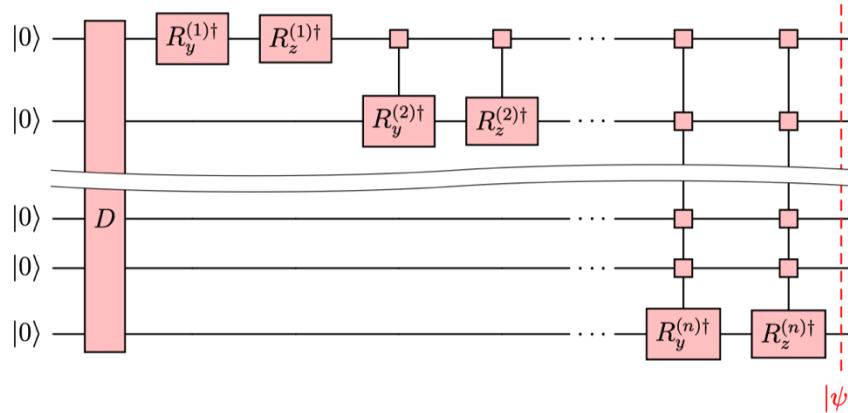
```
1 from numpy import linalg as LA
2 #np.random.seed(42)
3 nq = 4
4 ndim = 2**nq
5 state_vector = (2*np.random.rand(ndim)-1) * np.exp(1j*2*np.pi*np.random.rand(ndim))
6 state_vector /= LA.norm(state_vector)
7 mrot = rotate_to_vacuum_matrix(state_vector)
8 rot_vector = mrot.dot(state_vector)
9 back_vector = np.conjugate(mrot.T).dot(rot_vector)
```



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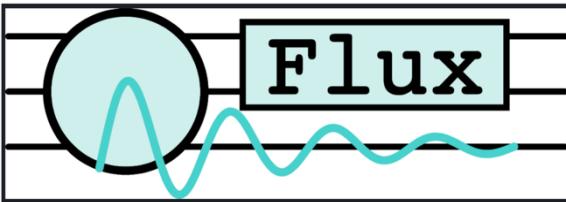
Recursive Disentanglement Algorithm

Script S.1.4: Recursive Quantum Multiplexor Transformation



What is the asymptotic cost? CNOT gates: $2^{n+1} - 2n$

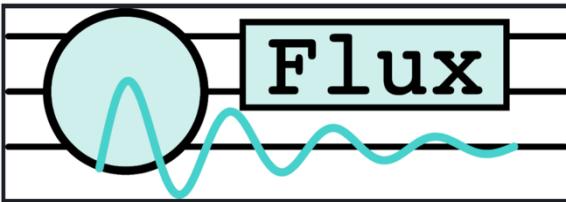
Single-qubit rotations: $O(2^n)$



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When to Use Multiplexors

- **Best for:**
 - Conceptual clarity
 - Debugging pipelines
 - Small n
- **Tradeoff:**
 - Higher CNOT count
 - Recursive, irregular structure



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PART B — UNIFORMLY CONTROLLED ROTATIONS (UCR)

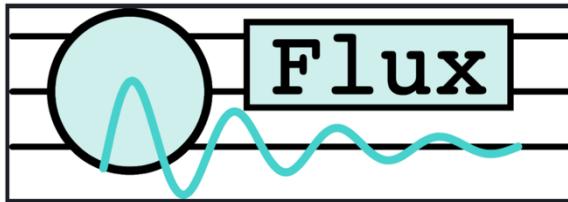
Motivation for UCRs

- Same goal: prepare arbitrary

$$|\psi\rangle = \left(|c_1|e^{i\omega_1}, |c_2|e^{i\omega_2}, \dots, |c_N|e^{i\omega_N} \right)^\top$$

- But:

- Non-recursive
- Regular layered structure
- Predictable gate counts



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Key Idea: Separate Phase and Amplitude

First, we make the state real and positive

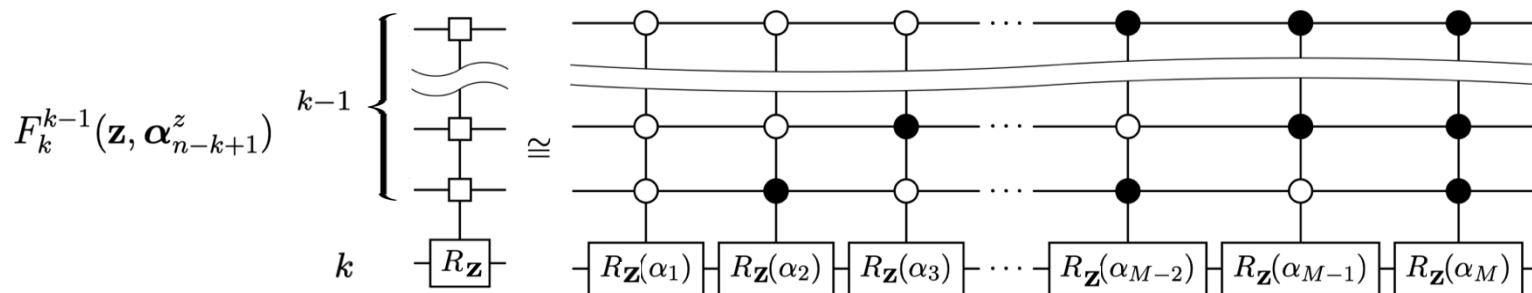
- **Step 1:** remove all relative phases with uniformly controlled z-rotations \mathcal{R}_z

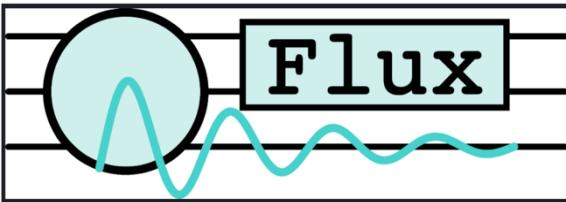
$$|\psi\rangle = (|c_1|e^{i\omega_1}, |c_2|e^{i\omega_2}, \dots, |c_N|e^{i\omega_N})^\top$$

Angles for phase equalization:

$$(\alpha_j)_k^{\mathbf{z}} = \sum_{l=1}^{2^{k-1}} \frac{\omega_{(2j-1)2^{k-1}+l} - \omega_{(j-1)2^k+l}}{2^{k-1}}, \quad j = 1, \dots, 2^{n-k}$$

$$\mathcal{R}_{\mathbf{z}} |\psi\rangle = e^{i\Omega}(|c_1|, |c_2|, \dots, |c_N|)^\top \quad \mathcal{R}_{\mathbf{z}} = \prod_{k=1} F_k^{k-1}(\mathbf{z}, \boldsymbol{\alpha}_{n-k+1}^z) \otimes I_{2^{n-k}}$$





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Key Idea: Separate Phase and Amplitude

Second, we move probability mass across basis states.

- **Step 2:** redistribute amplitudes with uniformly controlled y-rotations \mathcal{R}_y

$$\mathcal{R}_y = \prod_{k=1}^n F_k^{k-1}(\mathbf{y}, \boldsymbol{\alpha}_{n-k+1}^y) \otimes I_{2^{n-k}}$$

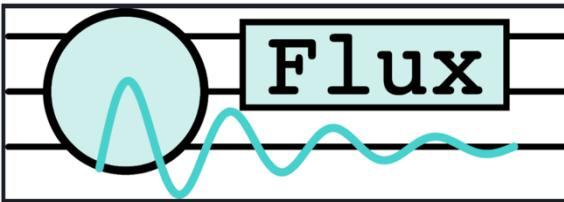
Angles for cancelling amplitudes: (with the least significant qubit set to 1)

$$(\alpha_j)_k^y = 2 \arcsin \left(\frac{\sqrt{\sum_{l=1}^{2^{k-1}} |c_{(2j-1)2^{k-1}+l}|^2}}{\sqrt{\sum_{l=1}^{2^k} |c_{(j-1)2^k+l}|^2}} \right)$$

$$\begin{aligned} \mathcal{R}_y \mathcal{R}_z |\psi\rangle &= e^{i\Phi} |0\dots0\rangle \\ |\psi\rangle &= \mathcal{R}_z^\dagger \mathcal{R}_y^\dagger e^{i\Phi} |0\dots0\rangle \end{aligned}$$

What is the asymptotic cost?

CNOT gates: $2^{n+2} - 4n - 4$; Single-qubit rotations: $2^{n+2} - 5$



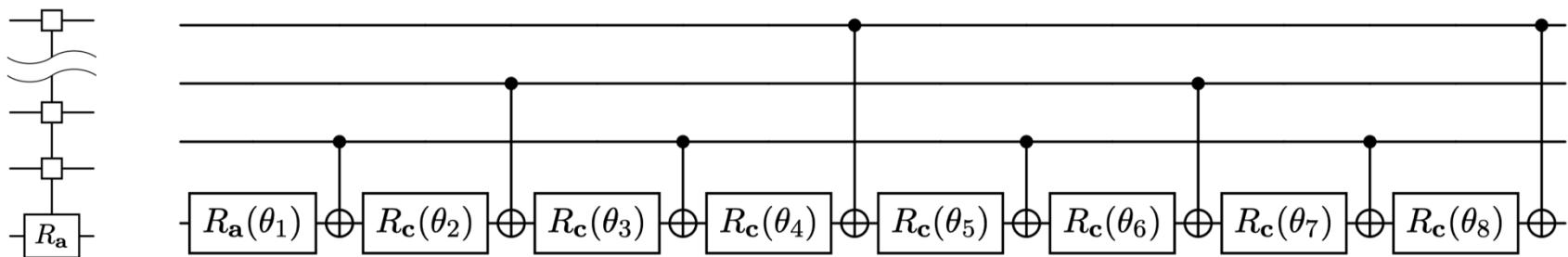
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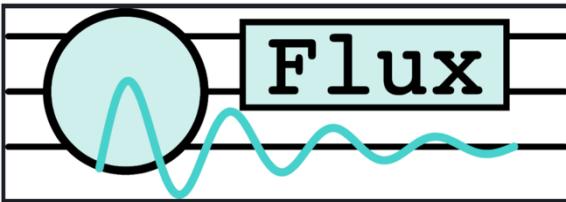
Gray Code

- Control patterns reordered by Gray code
- Consecutive gates differ by one control bit
- Enables:
 - Minimal CNOTs
 - Systematic cancellations

$$M_{ij} = 2^{-k}(-1)^{b_{j-1} \cdot g_{i-1}}$$

$$\begin{bmatrix} \theta_1 \\ \vdots \\ \theta_{2^k} \end{bmatrix} = \mathbf{M} \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_{2^k} \end{bmatrix}$$



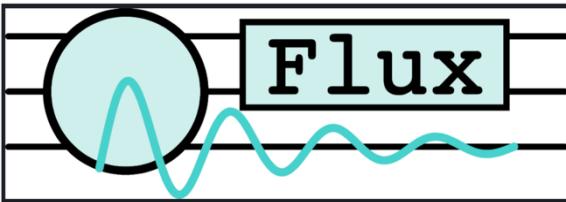


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When to Use UCRs

*In practice, this is your default state-preparation engine
Clean, compact, and hardware-friendly*

- Best for:
 - Scalable synthesis
 - Automation
 - Compiler pipelines
- Cost:
 - Fewer CNOTs than multiplexors
 - Analytic angles

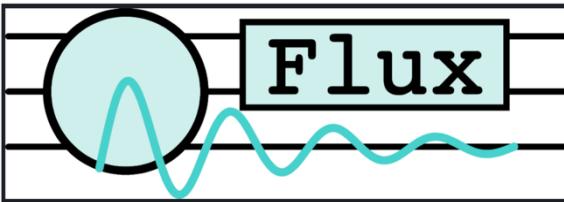


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PART C — GENERIC UNITARY DECOMPOSITION

Problem: Arbitrary Unitary Synthesis

- Input: $U \in \text{SU}(2^n)$
- Goal: decompose into 1- and 2-qubit gates
- Constraint: minimize CNOT count



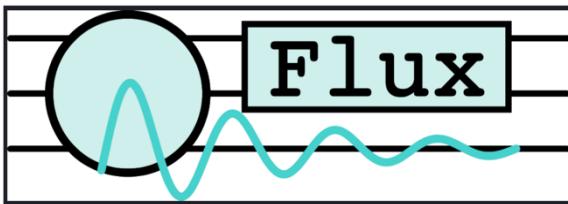
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Givens Rotations (QR Intuition)

- Classical QR → eliminate off-diagonal entries column by column
- Quantum version:
 - Two-level rotations
 - Cancel one element at a time

$${}^i\Gamma_{j,k} = \frac{1}{\sqrt{|U_{ji}|^2 + |U_{ki}|^2}} \begin{pmatrix} U_{ki}^* & U_{ji}^* \\ -U_{ji} & U_{ki} \end{pmatrix}$$

$${}^iG_{j,k} = \begin{pmatrix} \ddots & & & & & \\ & 1 & & & & \\ & \downarrow & & & & \\ & U_{ki}^* & & U_{ji}^* & & \\ & \cdots & 1 & \cdots & & \\ & \cdots & & \cdots & & \\ & -U_{ji} & & U_{ki} & & \\ & \cdots & & \cdots & 1 & \cdots \\ & & & & & \end{pmatrix} \begin{matrix} & k \\ \leftarrow & \\ j & \downarrow \\ & & \end{matrix}$$



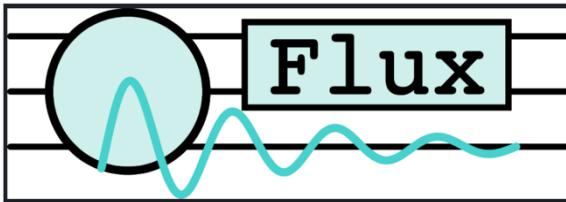
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Givens Rotations (QR Intuition)

Example: [4×4] unitary matrix U

$$\begin{aligned}
 U &\xrightarrow{e^{-i \arg(\det U)/4} I} \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & * \\ 0 & * & * & * \end{pmatrix} \xrightarrow{^1G_{4,3}} \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & * \\ 0 & * & * & * \end{pmatrix} \xrightarrow{^1G_{3,2}} \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ 0 & * & * & * \\ 0 & * & * & * \end{pmatrix} \xrightarrow{^1G_{2,1}} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & * & * & * \\ 0 & * & * & * \\ 0 & * & * & * \end{pmatrix} \\
 &\xrightarrow{^2G_{4,3}} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \end{pmatrix} \xrightarrow{^2G_{3,2}} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & * & * \\ 0 & 0 & * & * \end{pmatrix} \xrightarrow{^3G_{4,3}} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}
 \end{aligned}$$

$$U = \exp [i \arg (\det U) / 4] I ^1G_{4,3}^\dagger ^1G_{3,2}^\dagger ^1G_{2,1}^\dagger ^2G_{4,3}^\dagger ^2G_{3,2}^\dagger ^3G_{4,3}^\dagger$$



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Gray Code Makes It Implementable

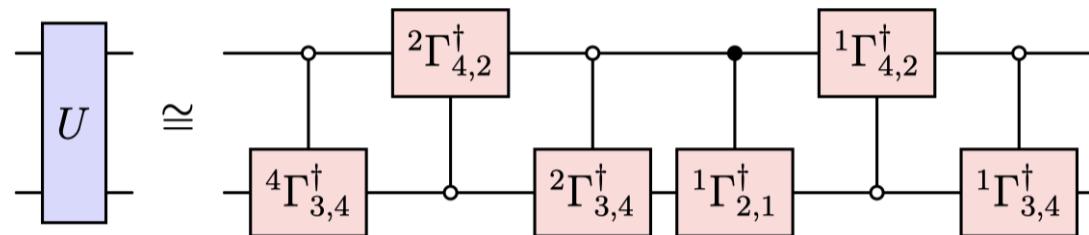
- Reorder basis states by Gray code
- Each rotation couples states differing by one qubit
- Implemented as multi-controlled single-qubit gates

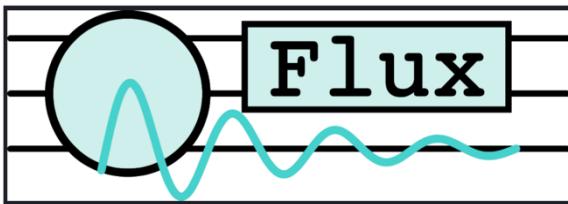
General formula:

$$\left[\prod_{i=1}^{N-1} \prod_{j=i+1}^N \gamma^{(N-i)} G_{\gamma(j), \gamma(j-1)} \right] \left(e^{-i \arg(\det U)/N} I \right) U = I \quad \gamma(\cdot) = \text{Gray-code permutation (i.e., the integer value of a Gray-coded bitstring)}$$

Example: 2 qubit unitary

[Part III.ipynb](#) (Scripts S.2.1–S.2.4)





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Column-by-Column Decomposition (CCD)

State preparation on every column

- Apply reverse state preparation to each column
- Protect previously fixed columns with extra controls

$$U = \mathcal{R}_0^\dagger \mathcal{R}_1^\dagger \cdots \mathcal{R}_{2^n-2}^\dagger D^\dagger$$

with $D = \mathcal{R}_{2^n-1}^\dagger$

1. Rotate its first column to $|0\rangle$ (state preparation):

$$\mathcal{R}_0 U_0 |0\rangle = |0\rangle \text{ with } U_0 := U$$

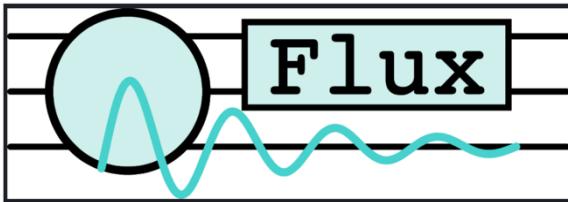
2. Rotate the second column of $U_1 := \mathcal{R}_0 U_0$ to $|1\rangle$:

$$\mathcal{R}_1 U_1 |1\rangle = |1\rangle \text{ with } \mathcal{R}_1 |0\rangle = |0\rangle$$

3. Rotate subsequent columns analogously:

$$\mathcal{R}_j U_j |j\rangle = |j\rangle \text{ with } U_{j+1} := \mathcal{R}_j U_j$$

$$\mathcal{R}_j |i\rangle = |i\rangle \text{ for all } i < j.$$



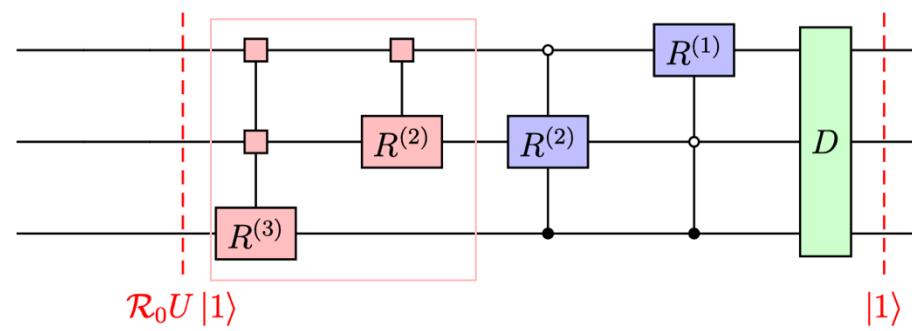
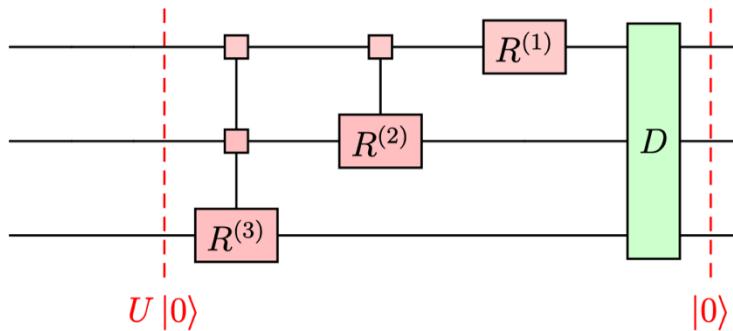
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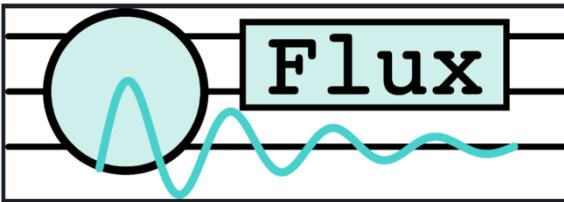
Column-by-Column Decomposition (CCD)

State preparation on every column

$$U |0\rangle = \begin{bmatrix} * \\ * \\ * \\ * \\ * \\ * \\ * \end{bmatrix} \rightarrow \begin{bmatrix} * \\ 0 \\ * \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} * \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} * \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathcal{R}_0 U |1\rangle = \begin{bmatrix} 0 \\ * \\ * \\ * \\ * \\ * \\ * \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ * \\ 0 \\ * \\ 0 \\ * \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ * \\ 0 \\ * \\ 0 \\ * \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ * \\ 0 \\ 0 \\ 0 \\ * \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ * \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$





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CSD & Quantum Shannon Decomposition

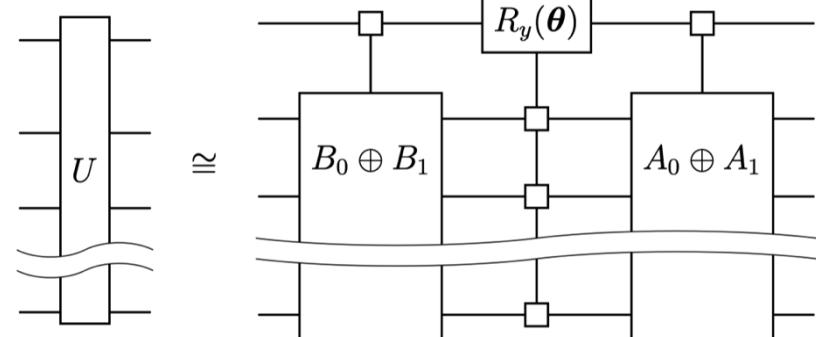
- Recursively split unitary into blocks
- Push all control into single-qubit multiplexed rotations

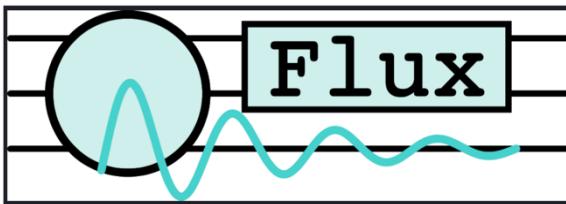
Cosine-Sine Decomposition (CSD): $U = \begin{pmatrix} U_{00} & U_{01} \\ U_{10} & U_{11} \end{pmatrix} = \underbrace{\begin{pmatrix} A_0 & 0 \\ 0 & A_1 \end{pmatrix}}_{A_0 \oplus A_1} \underbrace{\begin{pmatrix} C & -S \\ S & C \end{pmatrix}}_{\text{cos-sin block}} \underbrace{\begin{pmatrix} B_0 & 0 \\ 0 & B_1 \end{pmatrix}}_{B_0 \oplus B_1};$

$$C = \text{diag}(\cos \frac{\theta_1}{2}, \dots, \cos \frac{\theta_{2^n-1}}{2})$$

$$S = \text{diag}(\sin \frac{\theta_1}{2}, \dots, \sin \frac{\theta_{2^n-1}}{2})$$

Applying CSD recursively to $A_0 \oplus A_1$ and $B_0 \oplus B_1$





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CSD & Quantum Shannon Decomposition

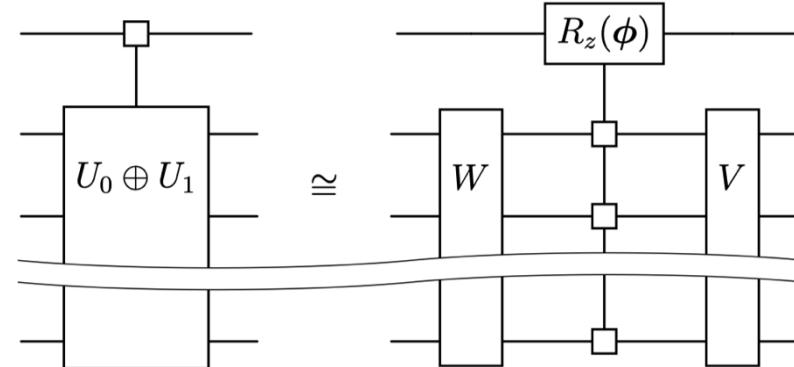
- Recursively split unitary into blocks
- Push all control into single-qubit multiplexed rotations

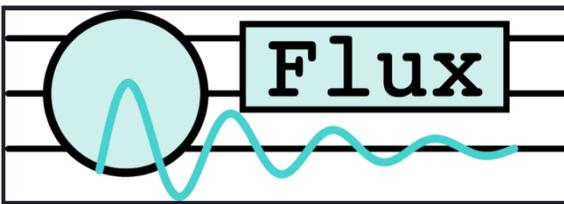
Quantum Shannon Decomposition (QSD):

$$U = \begin{pmatrix} V & 0 \\ 0 & V \end{pmatrix} \begin{pmatrix} D & 0 \\ 0 & D^\dagger \end{pmatrix} \begin{pmatrix} W & 0 \\ 0 & W \end{pmatrix}, \quad W = DV^\dagger U_1$$

$$U = \begin{pmatrix} U_0 & 0 \\ 0 & U_1 \end{pmatrix}$$

The middle block $D \oplus D^\dagger$ is a diagonal multiplexor and maps to a single multiplexed $R_z(\phi)$ on the most significant qubit

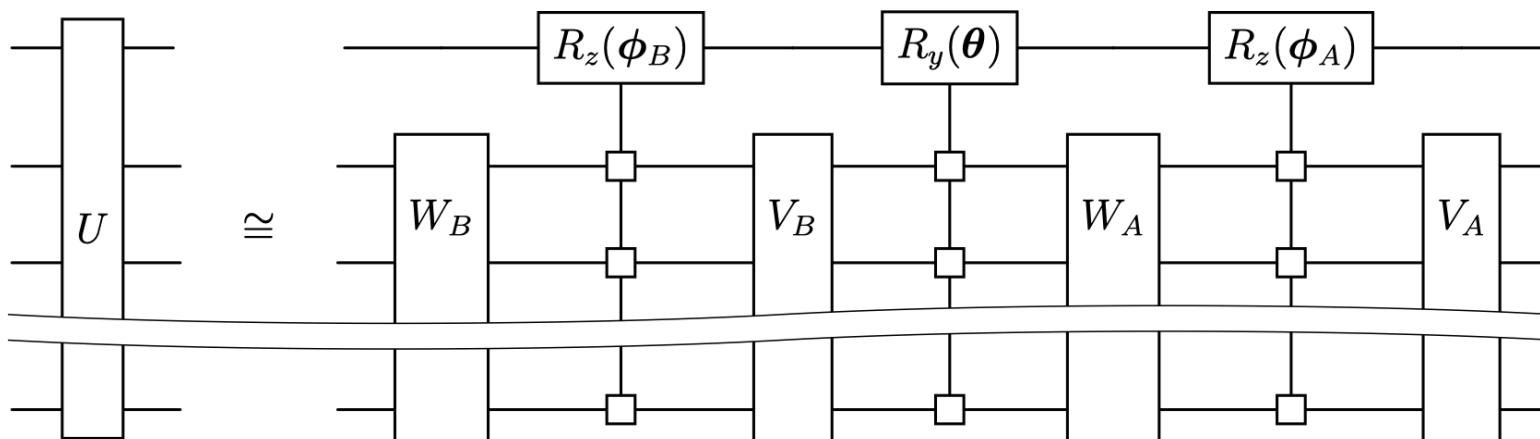




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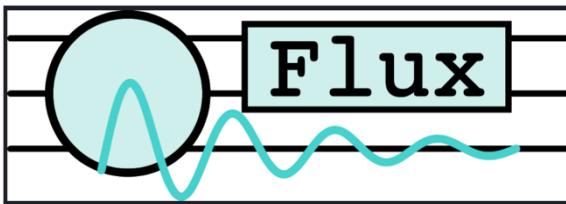
CSD + Quantum Shannon Decomposition

- Recursively split unitary into blocks
- Push all control into single-qubit multiplexed rotations



What is the asymptotic cost? CNOT gates: CSD: $O(4^n)$ (smaller constants)

Optimized QSD: $\frac{23}{48}4^n - \frac{3}{2}2^n + O(1)$

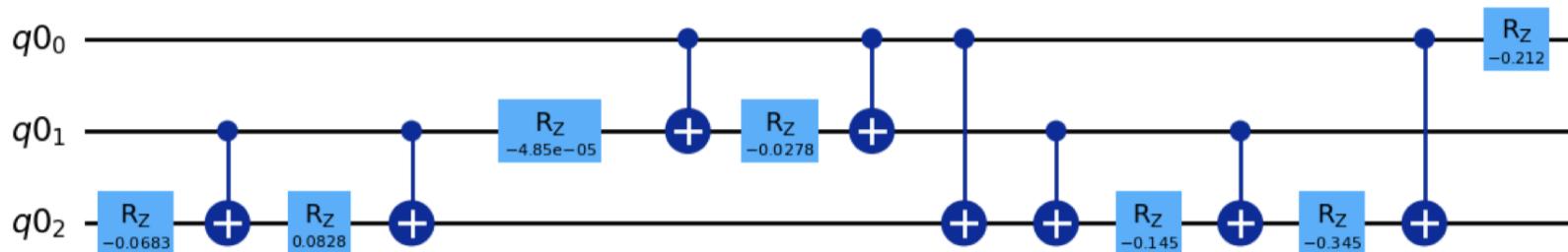


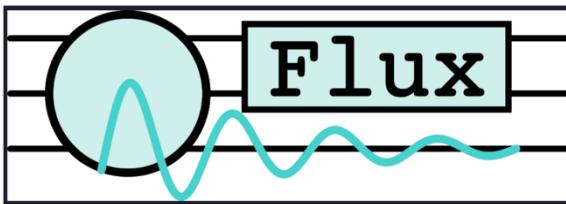
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PART D — DIAGONAL UNITARIES

Special Case: Diagonal Operators

- Arise in:
 - e^{-iHt}
 - Phase oracles
 - Lindblad dilation
- Best method:** Walsh decomposition
- [Part_III.ipynb](#) **Section S.3.6:** Walsh Synthesis of $e^{-iV(x_j)t}$ for a unitary of a Double-Well Potential
- Scripts S.3.1-5:** $U = \text{diag}(e^{if_k})$





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Walsh + Gray Code = NISQ-Optimal

- Expand diagonal operator in Walsh basis
- Order terms by Gray code
- Systematic CNOT cancellation
- Cost: $O(2^n)$ entangling gates

$$U = \prod_{j=0}^{2^n-1} e^{ia_j w_j}$$

Each factor $e^{ia_j w_j}$ is efficiently realized with one single-qubit Z-rotation and at most $2n$ CNOTs. Ordering terms by a Gray code cut two-qubit cost from naive $O(n2^n)$ to roughly $O(2^n)$ in practice

$$U = e^{iF}$$

$$F = \text{diag}(f_0, \dots, f_{2^n-1}) \text{ real}$$

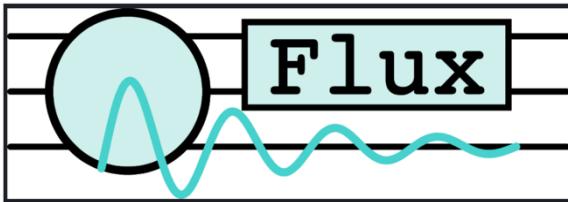
$$F = \sum_{j=0}^{2^n-1} a_j w_j$$

$$w_j := Z_1^{j_1} \otimes Z_2^{j_2} \otimes \cdots \otimes Z_n^{j_n}$$

$$a_j = 2^{-n} \text{Tr}(w_j F)$$

$j = (j_1 \dots j_n)_2$ is the *binary* label

$$j = \sum_{l=1}^n j_l 2^{n-l}$$



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Take-Home Messages

- State prep and unitary synthesis are the backbone of all simulations
- Operator structure determines optimal circuits
- QFlux provides:
 - Transparent algorithms
 - Predictable scaling
 - Reusable infrastructure