

Quantum Singular Value Transformation on NISQ Hardware

Yale Quantum Institute Visiting Student Researcher Program

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July 31, 2026



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Project Goals

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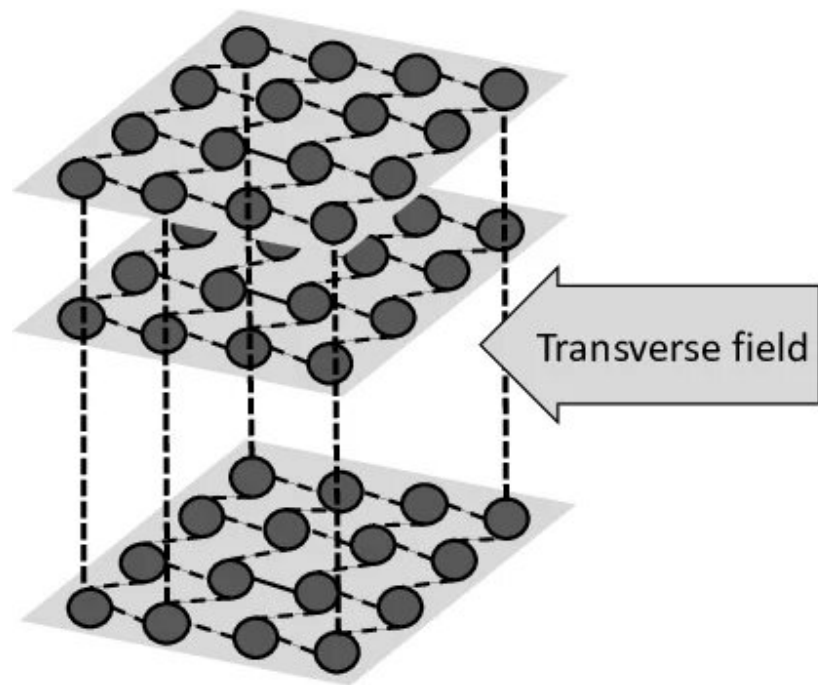
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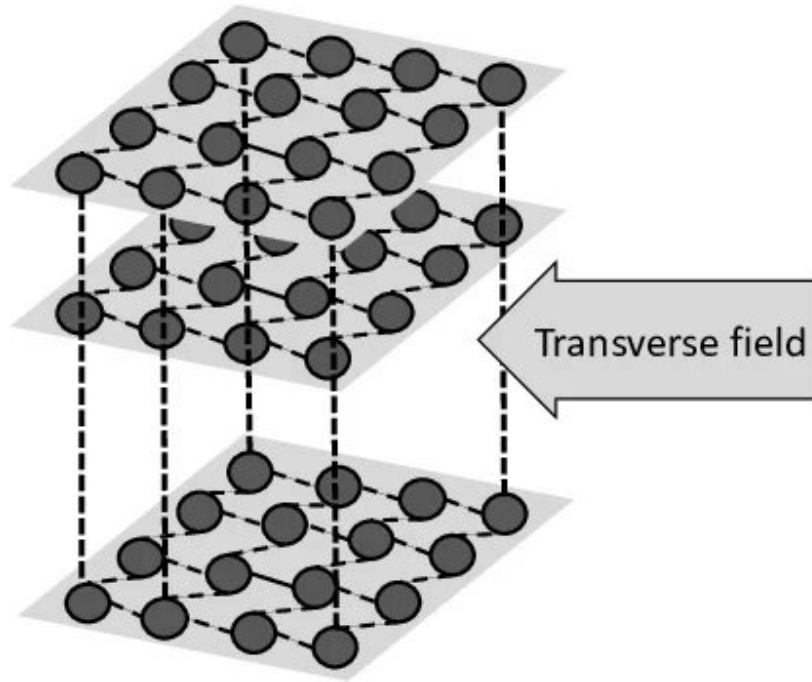
The Mixed Field Ising Hamiltonian:

$$H = -J \sum_{i=0}^{n-2} Z_i Z_{i+1} - \sum_{i=0}^{n-1} h_i X_i - m \sum_{i=0}^{n-1} Z_i$$

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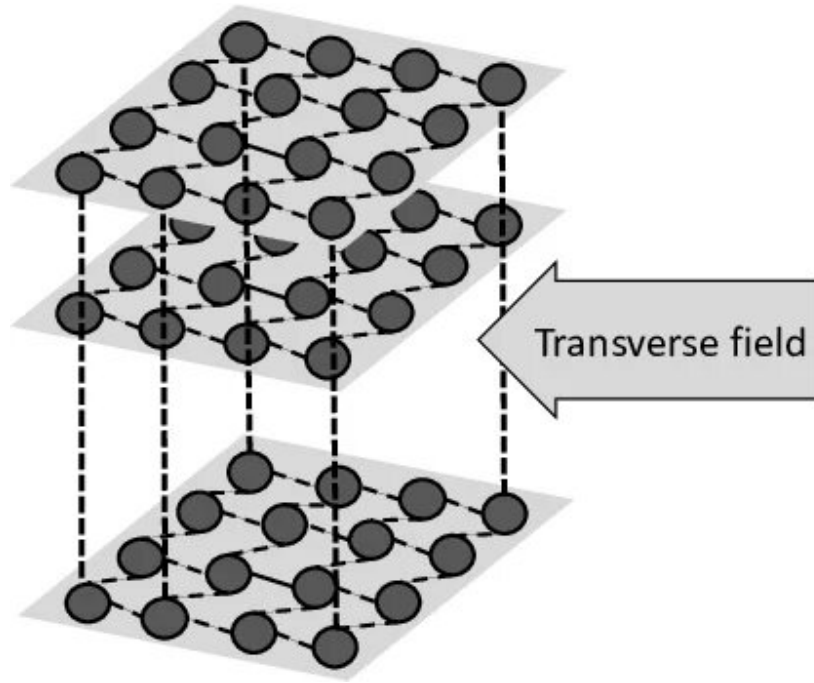
For $n = 3$:

$$H = -J Z_0 Z_1 - J Z_1 Z_2 - h_0 X_0 - h_1 X_1 - h_2 X_2 \\ - m(Z_0 + Z_1 + Z_2)$$

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We consider the $n = 3$ Ising spin chain with $h_i/J = -1.05$ for all i and $m/J = 0.5$ in our baseline Hamiltonian simulation.

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$$f(H) \text{ such that } f(x) = e^{-ixt}$$

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Quantum Singular Value Transformation (In Words)

The **quantum singular value transformation (QSVT)** is a technique that allows one to polynomially transform the singular values of a given matrix.

- For square matrices, the singular values are equal to the eigenvalues.
- Importantly, the QSVT has been shown to be an important primitive for many other quantum algorithms.
- **Three components:** Quantum Signal Processing, Block Encoding, Upscaling (Qubitization)

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Two big limitations with current hardware:

- Deeper circuits decohere states
- Cannot correct our errors at scale



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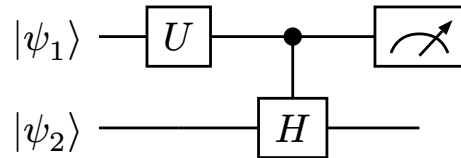
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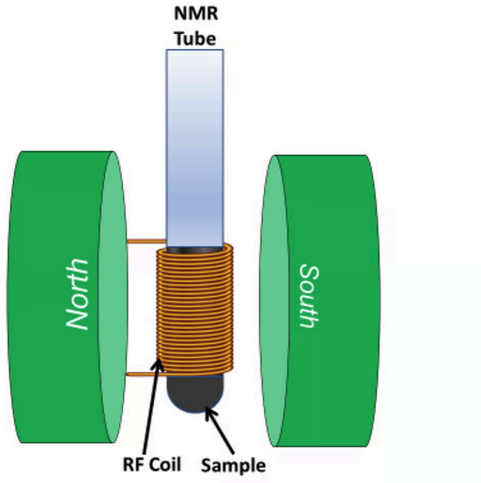
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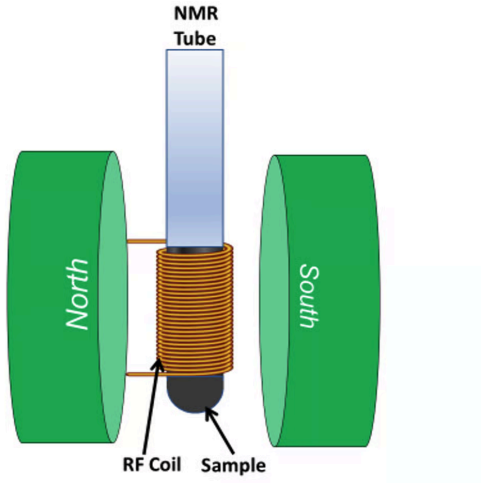
$$|\psi_1\rangle \otimes |\psi_2\rangle$$

Quantum Signal Processing

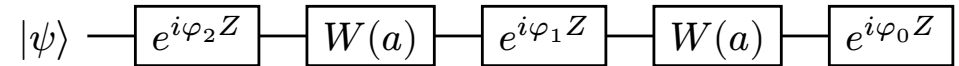
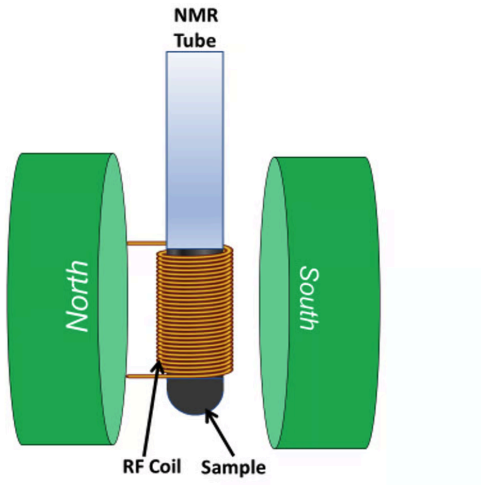
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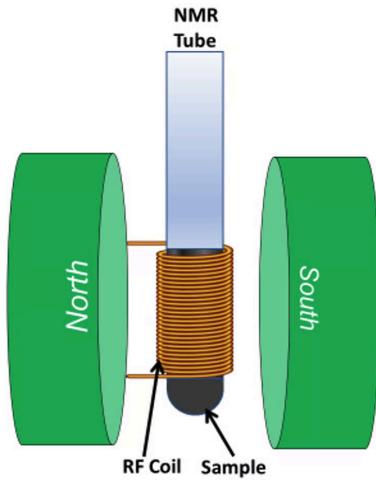
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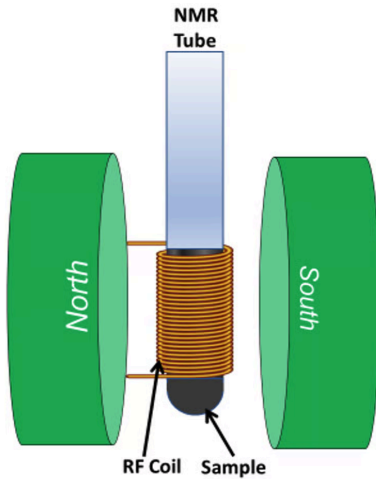
$$|\psi\rangle \rightarrow e^{i\varphi_2 Z} \rightarrow W(a) \rightarrow e^{i\varphi_1 Z} \rightarrow W(a) \rightarrow e^{i\varphi_0 Z}$$

Quantum Signal Processing (QSP)

Given a rotation matrix parameterized by a and a rotation matrix parameterized by a set of angles $\{\varphi_k\}$, we can interleave these operators to create a matrix $U_{\vec{\varphi}}$ that is a polynomial of a in the top left corner:

$$e^{i\varphi_0 Z} W(a) e^{i\varphi_1 Z} W(a) e^{i\varphi_2 Z} \dots W(a) e^{i\varphi_k Z} = \begin{pmatrix} P(a) & \cdot \\ \cdot & \cdot \end{pmatrix}$$

Quantum Signal Processing



$$e^{-iHt} = \cos(Ht) - i \sin(Ht)$$

$$\cos(Ht) \approx 1 - \frac{(Ht)^2}{2} + \frac{(Ht)^4}{24}$$

$$\sin(Ht) \approx Ht - \frac{(Ht)^3}{6} + \frac{(Ht)^5}{120}$$

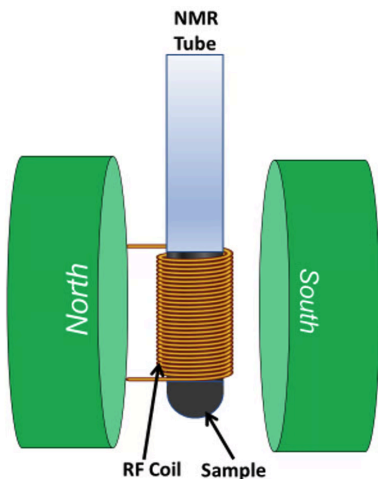
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Given a signal rotation operator $W(a) = \begin{pmatrix} a & i\sqrt{1-a^2} \\ i\sqrt{1-a^2} & a \end{pmatrix}$ and a signal processing rotation operator $e^{i\varphi Z}$, we may interleave these operators to create a matrix $U_{\vec{\varphi}}$ that can be expressed as a polynomial function of a (given various restrictions on the polynomials):

$$e^{i\varphi_0 Z} \prod_{k=1}^d W(a) e^{i\varphi_k Z} = \begin{pmatrix} P(a) & iQ(a)\sqrt{1-a^2} \\ iQ^*(a)\sqrt{1-a^2} & P^*(a) \end{pmatrix}$$

Block Encoding and Linear Combination of Unitaries

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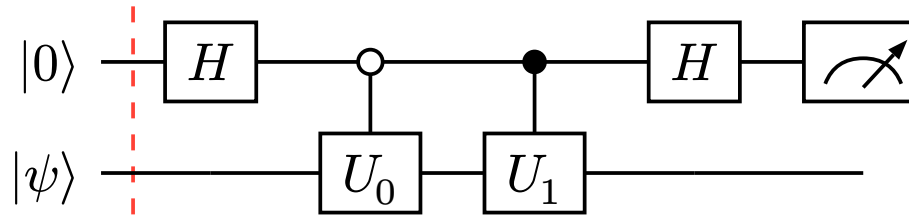
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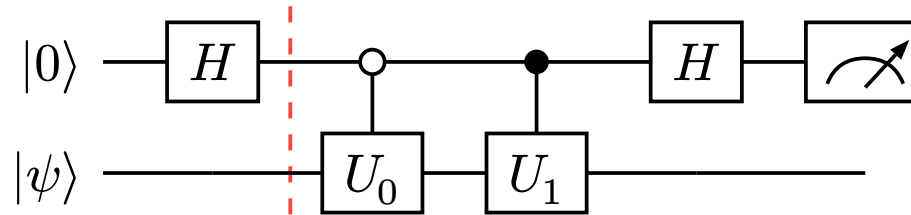
$$U = \begin{pmatrix} \mathbb{1} & 0 \\ 0 & \mathbb{1} \end{pmatrix}$$

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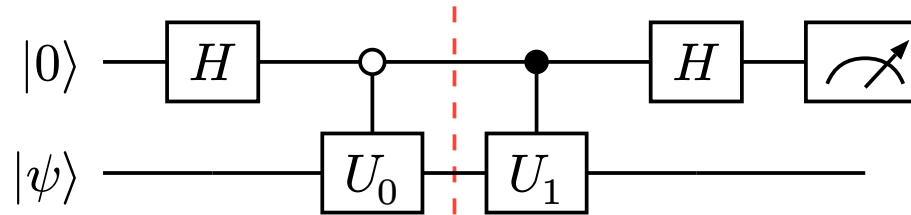
$$U = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

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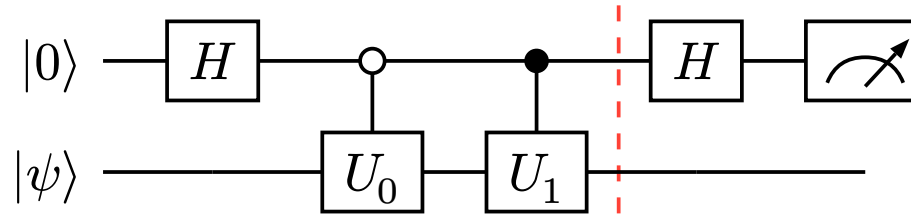


$$U = \frac{1}{\sqrt{2}} \begin{pmatrix} U_0 & U_0 \\ \mathbb{1} & -\mathbb{1} \end{pmatrix} \quad \frac{1}{\sqrt{2}} (|0\rangle \otimes U_0 |\psi\rangle + |1\rangle \otimes |\psi\rangle)$$

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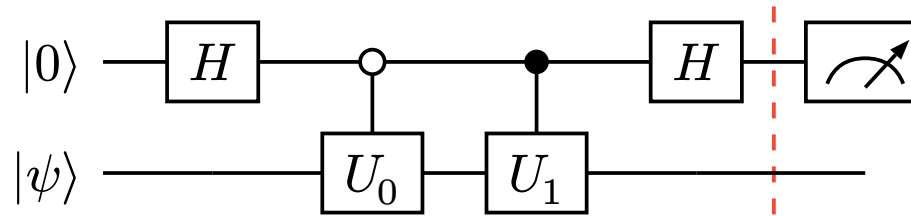
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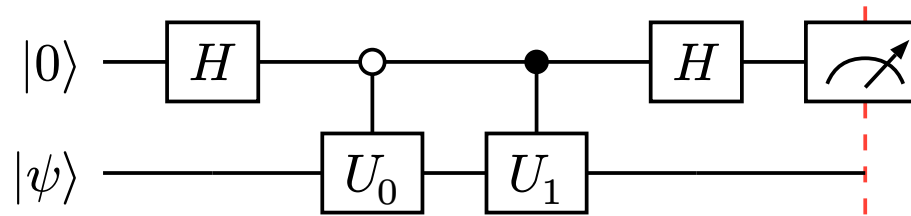
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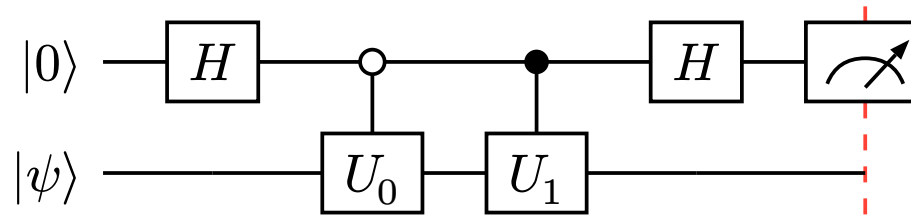
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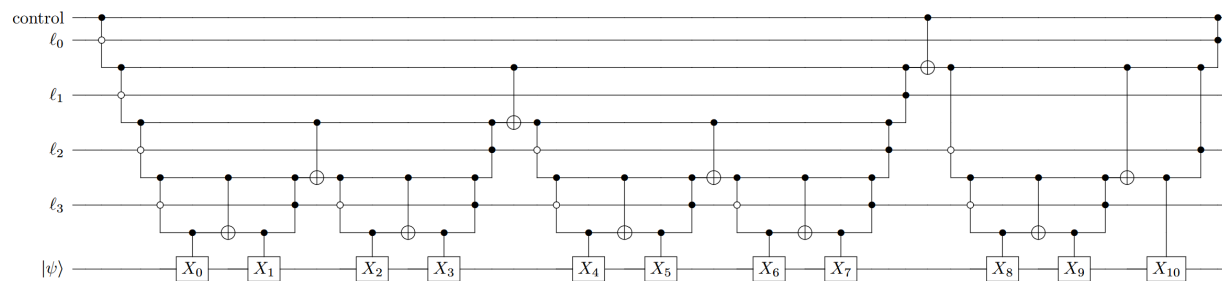
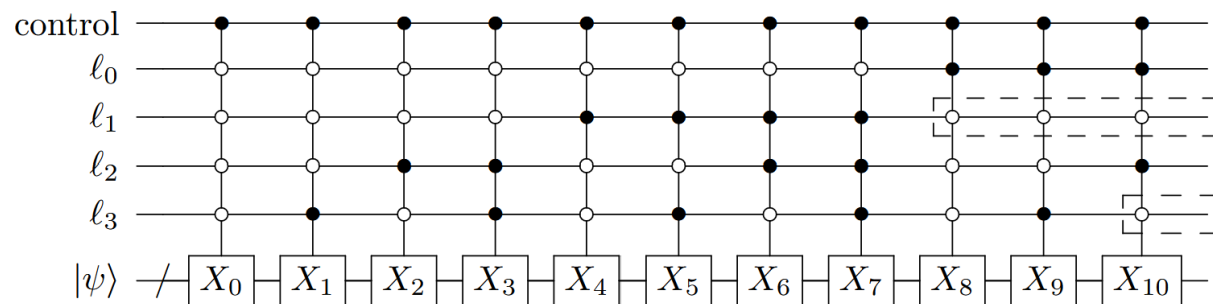


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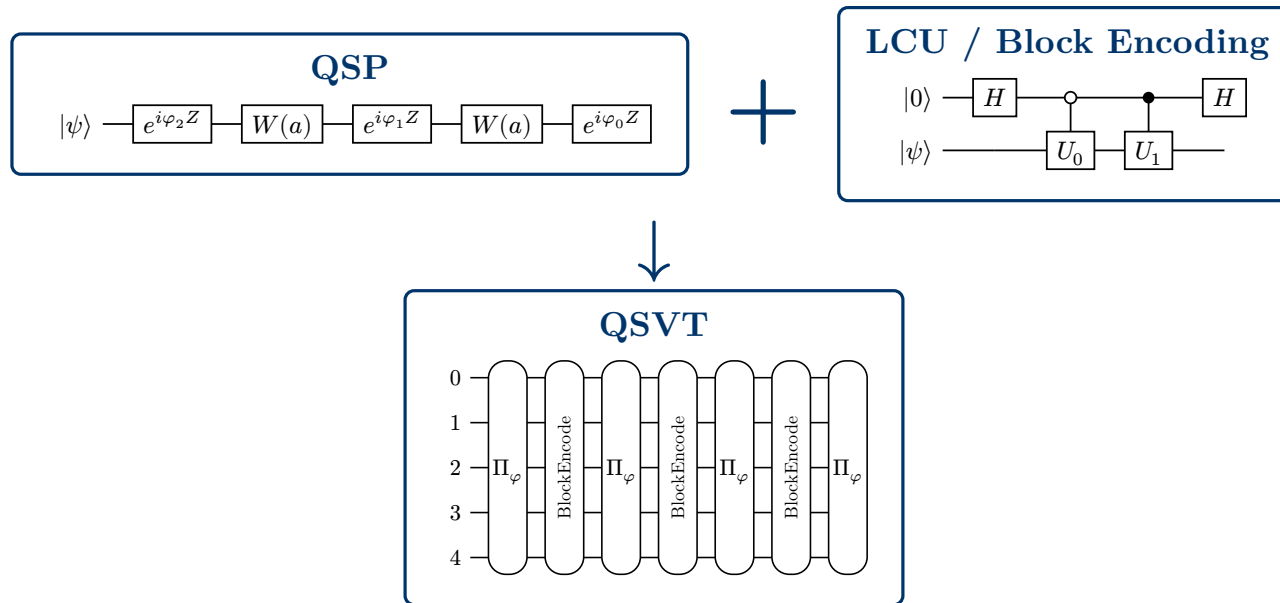
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Formal QSVT (Odd Parity)

Given a unitary U that block-encodes a matrix $A = \tilde{\Pi}U\Pi$ via projectors Π and $\tilde{\Pi}$, and projector-controlled phase shifts

$$\Pi_{\varphi} := e^{i\varphi(2\Pi-\mathbb{1})}, \quad \tilde{\Pi}_{\varphi} := e^{i\varphi(2\tilde{\Pi}-\mathbb{1})},$$

for odd degree d and phases $\vec{\varphi} = (\varphi_1, \dots, \varphi_d)$ the interleaved sequence

$$U_{\vec{\varphi}} = \tilde{\Pi}_{\varphi_1} U \prod_{k=1}^{\frac{d-1}{2}} (\Pi_{\varphi_{2k}} U^{\dagger} \tilde{\Pi}_{\varphi_{2k+1}} U)$$

implements an odd polynomial transform of the singular values of A :

$$\tilde{\Pi}U_{\vec{\varphi}}\Pi = P^{(SV)}(A).$$

Implementing the QSVT on NISQ Hardware

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Unoptimized resource estimation:

```
gateset = {"X", "Z", "S", "T", "Hadamard", "CNOT"}  
--- Resources: ---  
Total wires: 14  
  algorithmic wires: 8  
  allocated wires: 6  
    zero state: 6  
    any state: 0  
Total gates : 6.562E+4  
  'T': 3.665E+4,  
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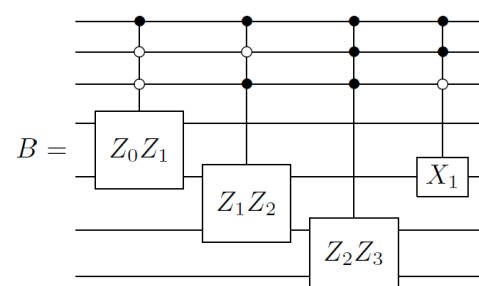
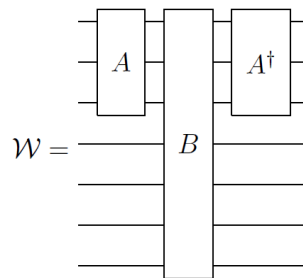
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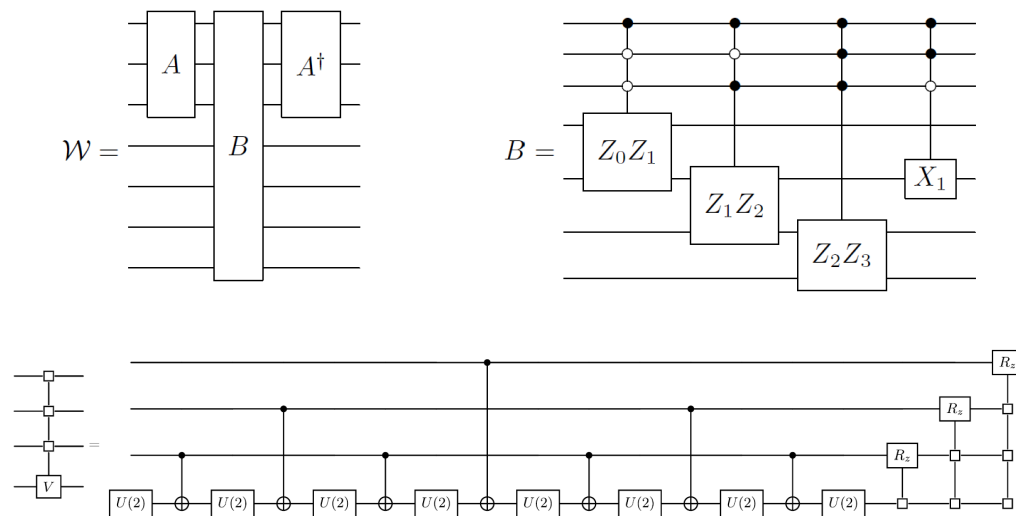


Figure 3. Decomposition for $M_n^C(V)$ where V is a list of $U(2)$ unitaries, $n=1$, and $C=3$. The open squares indicate multiplexed controls. (See figures 2 and 9).

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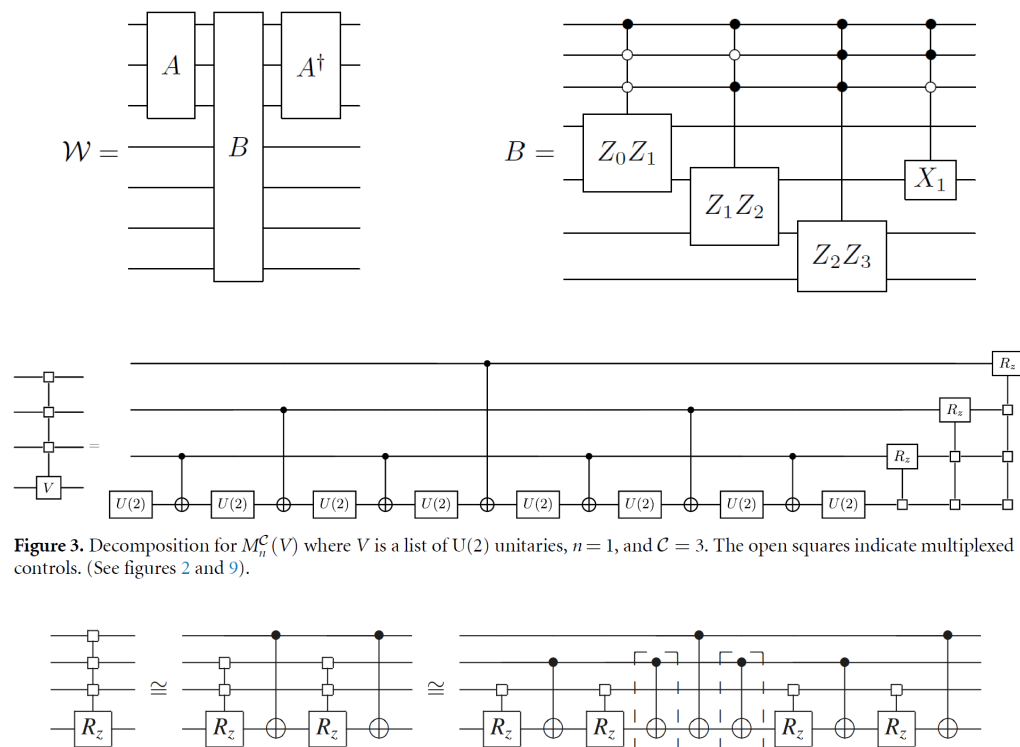


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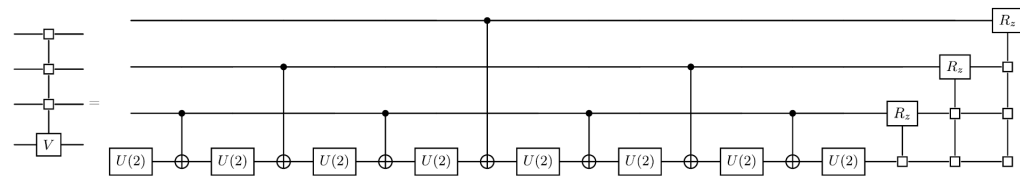


Figure 2: The recursive decomposition of a multiplexed R_z gate. The boxed CNOT gates may be canceled.

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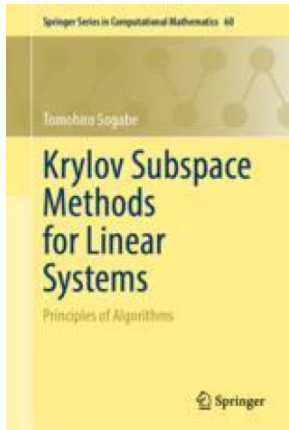
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Summary:

- Quantum Singular Value Transformation is a technique to polynomially transform the singular values of an arbitrary matrix
- We can formulate many algorithms in terms of the QSVT, but it is computationally expensive.
- Block encoding is the most expensive step of QSVT, but it can be simplified in many ways by looking for structure.

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Summary:

- Quantum Singular Value Transformation is a technique to polynomially transform the singular values of an arbitrary matrix
- We can formulate many algorithms in terms of the QSVT, but it is computationally expensive.
- Block encoding is the most expensive step of QSVT, but it can be simplified in many ways by looking for structure.

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Thank you for your time!

Questions?

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